

CustomerNYCStudentData

November 23, 2018

1 Machine Learning Engineer Nanodegree

1.1 Capstone Project

- Section ?? : Problem Definition Section
- Section ?? : Analysis Conducted
- Section ?? : Methodology Adopted
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Definition

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1.1.1 Project Overview: Creating Customer Segments from NYC School Data

In this project, we are considering data from Kaggle courtesy PASSNYC, a not-for-profit organization that facilitates a collective impact that is dedicated to broadening educational opportunities for New York City's talented and underserved students. PASSNYC uses public data to identify students within New York City's under-performing school districts and, through consulting and collaboration with partners, aims to increase the diversity of students taking the Specialized High School Admissions Test (SHSAT).

The objective is to assess the needs of students by using this data to quantify the challenges students face in taking the SHSAT. By segmenting and clustering, we can identify the schools where minority and underserved students stand to gain the most from services like after school programs, test preparation, mentoring, or resources for parents.

My personal motivation in selecting this domain is two-fold. I work for a company matching high school athletes to sports programs, so I wish to understand clustering better as applicable to school programs. Also, as an online student and micro-volunteer with school programs, it is of personal interest too.

1.1.2 Problem Statement:

In this project, the objective is to describe the variation in the different types of schools present in NYC. Based on demographic diversity or lack thereof other features of the school, the idea is to create categories of that represent subsets that together represent all the schools. Doing so will better equip these programs and how best to serve the schools.

To perform clustering on the data, we shall be using two techniques :Gaussian Mixture model and K-means clustering to create the clusters and decide on the optimal number of clusters.

1.1.3 Metrics:

Here, our solution is to determine the different segments present in the dataset. We will quantify the "goodness" of the clustering by calculating each data point's silhouette coefficient. The silhouette coefficient for a data point measures how similar it is to its assigned cluster from -1 (dissimilar) to 1 (similar). Calculating the mean silhouette coefficient will provide for a simple scoring method for our clustering and we shall take the best silhouette score of the different algorithms used. Silhouette score can be computed using sklearn library. Additionally, we will also verify the optimal number of clusters using the Elbow curve method.

1.2 Analysis

- Section ?? : Exploratory Analysis
- Section ?? : Vizs and Exploration
- Section ?? : Methods adopted
- Section ?? : Reference Model

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1.2.1 Data Exploration

The dataset are taken from Kaggle at: <https://www.kaggle.com/laiyipeng/target-schools-action-recommended-for-passnyc/data>

This is a dataset hosted by the City of New York and has demographic statistics broken down by school districts. It contains data about the percentage of students who belong to various races like White, Latino/Hispanic, African American, Pacific Islander or Alaskan, multi racial, economically disadvantaged. Their ELA and Math test scores are also provided, both averages as well as for each grade 3-8. We also have attributes such as Economic Need Index, Student Attendance Rate, Ratings on parameters such as Rigorous Instruction, Effective School Leadership, Trust, etc. By joining on the SHSAT dataset, we also get the corresponding values of how many students registered and then attempted the test, as well as how many were finally made an offer. Please find below the detailed exploration of the datasets.

```
In [577]: #Read and explore the datasets.
```

```
In [578]: # Import libraries necessary for this project
import numpy as np
import pandas as pd
import csv
```

```
# Import supplementary visualizations code visuals.py
#import visuals as vs
```

```
# Pretty display for notebooks
%matplotlib inline
```

```
df = pd.read_csv('2016 School Explorer.csv')
df.info()
df.head(3)
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1272 entries, 0 to 1271
Columns: 161 entries, Adjusted Grade to Grade 8 Math 4s - Economically Disadvantaged
dtypes: float64(5), int64(123), object(33)
memory usage: 1.6+ MB
```

```
Out[578]: Adjusted Grade New? Other Location Code in LCGMS School Name \
0          NaN NaN          NaN P.S. 015 ROBERTO CLEMENTE
1          NaN NaN          NaN P.S. 019 ASHER LEVY
2          NaN NaN          NaN P.S. 020 ANNA SILVER
```

```
SED Code Location Code District Latitude Longitude \
0 310100010015 01M015 1 40.721834 -73.978766
1 310100010019 01M019 1 40.729892 -73.984231
2 310100010020 01M020 1 40.721274 -73.986315
```

```
Address (Full) \
0 333 E 4TH ST NEW YORK, NY 10009
1 185 1ST AVE NEW YORK, NY 10003
2 166 ESSEX ST NEW YORK, NY 10002
```

```
... \
0 ...
1 ...
2 ...
```

```
Grade 8 Math - All Students Tested Grade 8 Math 4s - All Students \
0 0 0
1 0 0
2 0 0
```

```
Grade 8 Math 4s - American Indian or Alaska Native \
0 0
1 0
2 0
```

	Grade 8 Math 4s - Black or African American \
0	0
1	0
2	0

	Grade 8 Math 4s - Hispanic or Latino \
0	0
1	0
2	0

	Grade 8 Math 4s - Asian or Pacific Islander	Grade 8 Math 4s - White \
0	0	0
1	0	0
2	0	0

	Grade 8 Math 4s - Multiracial	Grade 8 Math 4s - Limited English Proficient \
0	0	0
1	0	0
2	0	0

	Grade 8 Math 4s - Economically Disadvantaged
0	0
1	0
2	0

[3 rows x 161 columns]

```
In [579]: df1 = pd.read_csv('D5 SHSAT Registrations and Testers.csv')
df1.info()
df1.head(3)
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 140 entries, 0 to 139
Data columns (total 7 columns):
DBN                                140 non-null object
School name                        140 non-null object
Year of SHST                       140 non-null int64
Grade level                        140 non-null int64
Enrollment on 10/31               140 non-null int64
Number of students who registered for the SHSAT  140 non-null int64
Number of students who took the SHSAT          140 non-null int64
dtypes: int64(5), object(2)
memory usage: 7.7+ KB
```

```
Out [579]:
```

	DBN	School name	Year of SHST	Grade level \
0	05M046	P.S. 046 Arthur Tappan	2013	8
1	05M046	P.S. 046 Arthur Tappan	2014	8

2	05M046	P.S. 046 Arthur Tappan	2015	8
---	--------	------------------------	------	---

	Enrollment on 10/31	Number of students who registered for the SHSAT	\
0	91		31
1	95		26
2	73		21

	Number of students who took the SHSAT
0	14
1	7
2	10

```
In [580]: #2017-2018_SHSAT_Admissions_Test_Offers_By_Sending_School.csv
df3 = pd.read_csv('2017-2018_SHSAT_Admissions_Test_Offers_By_Sending_School.csv')
df3.info()
df3.head(3)
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 594 entries, 0 to 593
Data columns (total 5 columns):
Feeder School DBN          594 non-null object
Feeder School Name         594 non-null object
Count of Students in HS Admissions  594 non-null int64
Count of Testers           594 non-null int64
Count of Offers            594 non-null int64
dtypes: int64(3), object(2)
memory usage: 23.3+ KB
```

```
Out[580]:
```

	Feeder School DBN	Feeder School Name	\
0	01M034	P.S. 034 FRANKLIN D. ROOSEVELT	
1	01M140	P.S. 140 NATHAN STRAUS	
2	01M184	P.S. 184M SHUANG WEN	

	Count of Students in HS Admissions	Count of Testers	Count of Offers
0	58	6	5
1	67	6	5
2	88	67	23

```
In [581]: dfx= df.merge(df1, how='left', right_on='DBN' , left_on='Location Code')
dfx.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1364 entries, 0 to 1363
Columns: 168 entries, Adjusted Grade to Number of students who took the SHSAT
dtypes: float64(10), int64(123), object(35)
memory usage: 1.8+ MB
```

```
In [582]: dfx.shape
```

```
Out[582]: (1364, 168)
```

```
In [583]: dfy = df.merge(df1, how='left', right_on='DBN' , left_on='Location Code')
dfy.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Int64Index: 1364 entries, 0 to 1363
```

```
Columns: 168 entries, Adjusted Grade to Number of students who took the SHSAT
```

```
dtypes: float64(10), int64(123), object(35)
```

```
memory usage: 1.8+ MB
```

```
In [584]: dfy.head()
```

```
Out[584]: Adjusted Grade New? Other Location Code in LCGMS \
```

0	NaN	NaN	NaN
1	NaN	NaN	NaN
2	NaN	NaN	NaN
3	NaN	NaN	NaN
4	NaN	NaN	NaN

	School Name	SED Code	Location Code	District	\
0	P.S. 015 ROBERTO CLEMENTE	310100010015	01M015	1	
1	P.S. 019 ASHER LEVY	310100010019	01M019	1	
2	P.S. 020 ANNA SILVER	310100010020	01M020	1	
3	P.S. 034 FRANKLIN D. ROOSEVELT	310100010034	01M034	1	
4	THE STAR ACADEMY - P.S.63	310100010063	01M063	1	

	Latitude	Longitude	Address (Full)	\
0	40.721834	-73.978766	333 E 4TH ST NEW YORK, NY 10009	
1	40.729892	-73.984231	185 1ST AVE NEW YORK, NY 10003	
2	40.721274	-73.986315	166 ESSEX ST NEW YORK, NY 10002	
3	40.726147	-73.975043	730 E 12TH ST NEW YORK, NY 10009	
4	40.724404	-73.986360	121 E 3RD ST NEW YORK, NY 10009	

	...	Grade 8 Math 4s - Multiracial	\
0	...	0	
1	...	0	
2	...	0	
3	...	0	
4	...	0	

	Grade 8 Math 4s - Limited English Proficient	\
0	0	
1	0	
2	0	
3	0	

4

0

	Grade 8 Math 4s - Economically Disadvantaged	DBN	School name	Year of SHST	\
0	0	NaN	NaN	NaN	
1	0	NaN	NaN	NaN	
2	0	NaN	NaN	NaN	
3	0	NaN	NaN	NaN	
4	0	NaN	NaN	NaN	

	Grade level Enrollment on 10/31	\
0	NaN	NaN
1	NaN	NaN
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN

	Number of students who registered for the SHSAT	\
0	NaN	
1	NaN	
2	NaN	
3	NaN	
4	NaN	

	Number of students who took the SHSAT
0	NaN
1	NaN
2	NaN
3	NaN
4	NaN

[5 rows x 168 columns]

In [585]: # Display a description of the dataset
display(dfy.describe())

	SED Code	District	Latitude	Longitude	Zip	\
count	1.364000e+03	1364.000000	1364.000000	1364.000000	1364.000000	
mean	3.274443e+11	15.384164	40.739964	-73.920300	10762.730938	
std	1.265045e+10	9.354700	0.086044	0.078165	548.074632	
min	3.075000e+11	1.000000	40.507803	-74.244025	10001.000000	
25%	3.207000e+11	7.000000	40.672706	-73.955847	10304.000000	
50%	3.314000e+11	14.000000	40.736511	-73.925881	10738.000000	
75%	3.332000e+11	24.000000	40.816396	-73.884761	11228.000000	
max	3.531009e+11	32.000000	40.903455	-73.708920	11694.000000	

	Economic Need Index	Average ELA Proficiency	Average Math Proficiency	\
count	1339.000000	1309.000000	1309.000000	
mean	0.677808	2.530527	2.667540	

std	0.205944	0.361546	0.473053
min	0.049000	1.810000	1.830000
25%	0.559500	2.260000	2.290000
50%	0.737000	2.450000	2.580000
75%	0.839000	2.740000	2.990000
max	0.957000	3.930000	4.200000

	Grade 3 ELA - All Students Tested	Grade 3 ELA 4s - All Students \
count	1364.000000	1364.000000
mean	58.475806	4.734604
std	57.315286	8.084927
min	0.000000	0.000000
25%	0.000000	0.000000
50%	52.000000	1.000000
75%	93.000000	6.000000
max	356.000000	55.000000

	...	\
count	...	
mean	...	
std	...	
min	...	
25%	...	
50%	...	
75%	...	
max	...	

	Grade 8 Math 4s - Asian or Pacific Islander	Grade 8 Math 4s - White \
count	1364.000000	1364.000000
mean	1.850440	0.905425
std	12.410368	6.648429
min	0.000000	0.000000
25%	0.000000	0.000000
50%	0.000000	0.000000
75%	0.000000	0.000000
max	246.000000	126.000000

	Grade 8 Math 4s - Multiracial \
count	1364.000000
mean	0.002199
std	0.081230
min	0.000000
25%	0.000000
50%	0.000000
75%	0.000000
max	3.000000

Grade 8 Math 4s - Limited English Proficient \

count	1364.000000
mean	0.148827
std	1.276455
min	0.000000
25%	0.000000
50%	0.000000
75%	0.000000
max	33.000000

	Grade 8 Math 4s - Economically Disadvantaged	Year of SHST \
count	1364.000000	113.000000
mean	3.000000	2014.601770
std	12.312154	1.122337
min	0.000000	2013.000000
25%	0.000000	2014.000000
50%	0.000000	2015.000000
75%	1.000000	2016.000000
max	196.000000	2016.000000

	Grade level	Enrollment on 10/31 \
count	113.000000	113.000000
mean	8.309735	93.486726
std	0.464444	48.200310
min	8.000000	35.000000
25%	8.000000	66.000000
50%	8.000000	86.000000
75%	9.000000	110.000000
max	9.000000	344.000000

	Number of students who registered for the SHSAT \
count	113.000000
mean	22.716814
std	24.419865
min	0.000000
25%	4.000000
50%	17.000000
75%	30.000000
max	118.000000

	Number of students who took the SHSAT
count	113.000000
mean	11.442478
std	10.398264
min	0.000000
25%	3.000000
50%	10.000000
75%	17.000000
max	45.000000

[8 rows x 133 columns]

```
In [586]: # Display a description of the dataset
display(dfx.describe())
```

	SED Code	District	Latitude	Longitude	Zip \
count	1.364000e+03	1364.000000	1364.000000	1364.000000	1364.000000
mean	3.274443e+11	15.384164	40.739964	-73.920300	10762.730938
std	1.265045e+10	9.354700	0.086044	0.078165	548.074632
min	3.075000e+11	1.000000	40.507803	-74.244025	10001.000000
25%	3.207000e+11	7.000000	40.672706	-73.955847	10304.000000
50%	3.314000e+11	14.000000	40.736511	-73.925881	10738.000000
75%	3.332000e+11	24.000000	40.816396	-73.884761	11228.000000
max	3.531009e+11	32.000000	40.903455	-73.708920	11694.000000

	Economic Need Index	Average ELA Proficiency	Average Math Proficiency \
count	1339.000000	1309.000000	1309.000000
mean	0.677808	2.530527	2.667540
std	0.205944	0.361546	0.473053
min	0.049000	1.810000	1.830000
25%	0.559500	2.260000	2.290000
50%	0.737000	2.450000	2.580000
75%	0.839000	2.740000	2.990000
max	0.957000	3.930000	4.200000

	Grade 3 ELA - All Students Tested	Grade 3 ELA 4s - All Students \
count	1364.000000	1364.000000
mean	58.475806	4.734604
std	57.315286	8.084927
min	0.000000	0.000000
25%	0.000000	0.000000
50%	52.000000	1.000000
75%	93.000000	6.000000
max	356.000000	55.000000

	...	\
count	...	
mean	...	
std	...	
min	...	
25%	...	
50%	...	
75%	...	
max	...	

	Grade 8 Math 4s - Asian or Pacific Islander	Grade 8 Math 4s - White \
--	---	---------------------------

count	1364.000000	1364.000000
mean	1.850440	0.905425
std	12.410368	6.648429
min	0.000000	0.000000
25%	0.000000	0.000000
50%	0.000000	0.000000
75%	0.000000	0.000000
max	246.000000	126.000000

Grade 8 Math 4s - Multiracial \

count	1364.000000
mean	0.002199
std	0.081230
min	0.000000
25%	0.000000
50%	0.000000
75%	0.000000
max	3.000000

Grade 8 Math 4s - Limited English Proficient \

count	1364.000000
mean	0.148827
std	1.276455
min	0.000000
25%	0.000000
50%	0.000000
75%	0.000000
max	33.000000

Grade 8 Math 4s - Economically Disadvantaged Year of SHST \

count	1364.000000	113.000000
mean	3.000000	2014.601770
std	12.312154	1.122337
min	0.000000	2013.000000
25%	0.000000	2014.000000
50%	0.000000	2015.000000
75%	1.000000	2016.000000
max	196.000000	2016.000000

Grade level Enrollment on 10/31 \

count	113.000000	113.000000
mean	8.309735	93.486726
std	0.464444	48.200310
min	8.000000	35.000000
25%	8.000000	66.000000
50%	8.000000	86.000000
75%	9.000000	110.000000
max	9.000000	344.000000

	Number of students who registered for the SHSAT \
count	113.000000
mean	22.716814
std	24.419865
min	0.000000
25%	4.000000
50%	17.000000
75%	30.000000
max	118.000000

	Number of students who took the SHSAT
count	113.000000
mean	11.442478
std	10.398264
min	0.000000
25%	3.000000
50%	10.000000
75%	17.000000
max	45.000000

[8 rows x 133 columns]

```
In [587]: dfy['AllTested'] = dfy[['Grade 3 ELA - All Students Tested',
    'Grade 4 ELA - All Students Tested',
    'Grade 5 ELA - All Students Tested',
    'Grade 6 ELA - All Students Tested',
    'Grade 7 ELA - All Students Tested',
    'Grade 8 ELA - All Students Tested']].mean(axis=1)
```

```
dfy['All4'] = dfy[['Grade 3 ELA 4s - All Students',
    'Grade 4 ELA 4s - All Students',
    'Grade 5 ELA 4s - All Students',
    'Grade 6 ELA 4s - All Students',
    'Grade 7 ELA 4s - All Students',
    'Grade 8 ELA 4s - All Students']].mean(axis=1)
```

```
dfy['Native4'] = dfy[['Grade 3 ELA 4s - American Indian or Alaska Native',
    'Grade 4 ELA 4s - American Indian or Alaska Native',
    'Grade 5 ELA 4s - American Indian or Alaska Native',
    'Grade 6 ELA 4s - American Indian or Alaska Native',
    'Grade 7 ELA 4s - American Indian or Alaska Native',
    'Grade 8 ELA 4s - American Indian or Alaska Native']].mean(axis=1)
```

```
dfy['AfricanAmerican4'] = dfy[['Grade 3 ELA 4s - Black or African American',
    'Grade 4 ELA 4s - Black or African American',
    'Grade 5 ELA 4s - Black or African American',
```

```

        'Grade 6 ELA 4s - Black or African American',
        'Grade 7 ELA 4s - Black or African American',
        'Grade 8 ELA 4s - Black or African American']] .mean(axis=1)

dfy['Latino4'] = dfy[['Grade 3 ELA 4s - Hispanic or Latino',
                    'Grade 4 ELA 4s - Hispanic or Latino',
                    'Grade 5 ELA 4s - Hispanic or Latino',
                    'Grade 6 ELA 4s - Hispanic or Latino',
                    'Grade 7 ELA 4s - Hispanic or Latino',
                    'Grade 8 ELA 4s - Hispanic or Latino']] .mean(axis=1)

dfy['Islander4'] = dfy[['Grade 3 ELA 4s - Asian or Pacific Islander',
                    'Grade 4 ELA 4s - Asian or Pacific Islander',
                    'Grade 5 ELA 4s - Asian or Pacific Islander',
                    'Grade 6 ELA 4s - Asian or Pacific Islander',
                    'Grade 7 ELA 4s - Asian or Pacific Islander',
                    'Grade 8 ELA 4s - Asian or Pacific Islander']] .mean(axis=1)

dfy['White4'] = dfy[['Grade 3 ELA 4s - White',
                    'Grade 4 ELA 4s - White',
                    'Grade 5 ELA 4s - White',
                    'Grade 6 ELA 4s - White',
                    'Grade 7 ELA 4s - White',
                    'Grade 8 ELA 4s - White']] .mean(axis=1)

dfy['Multiracial4'] = dfy[['Grade 3 ELA 4s - Multiracial',
                    'Grade 4 ELA 4s - Multiracial',
                    'Grade 5 ELA 4s - Multiracial',
                    'Grade 6 ELA 4s - Multiracial',
                    'Grade 7 ELA 4s - Multiracial',
                    'Grade 8 ELA 4s - Multiracial']] .mean(axis=1)

dfy['LimitedEnglish4'] = dfy[['Grade 3 ELA 4s - Limited English Proficient',
                    'Grade 4 ELA 4s - Limited English Proficient',
                    'Grade 5 ELA 4s - Limited English Proficient',
                    'Grade 6 ELA 4s - Limited English Proficient',
                    'Grade 7 ELA 4s - Limited English Proficient',
                    'Grade 8 ELA 4s - Limited English Proficient']] .mean(axis=1)

dfy['Disadv4'] = dfy[['Grade 3 ELA 4s - Economically Disadvantaged',
                    'Grade 4 ELA 4s - Economically Disadvantaged',
                    'Grade 5 ELA 4s - Economically Disadvantaged',
                    'Grade 6 ELA 4s - Economically Disadvantaged',
                    'Grade 7 ELA 4s - Economically Disadvantaged',
                    'Grade 8 ELA 4s - Economically Disadvantaged']] .mean(axis=1)

dfy['AllMath4Tested'] = dfy[['Grade 3 Math - All Students tested',
                    'Grade 4 Math - All Students Tested',

```

```

        'Grade 5 Math - All Students Tested',
        'Grade 6 Math - All Students Tested',
        'Grade 7 Math - All Students Tested',
        'Grade 8 Math - All Students Tested']] .mean(axis=1)

dfy['AllMath4'] = dfy[['Grade 3 Math 4s - All Students',
        'Grade 4 Math 4s - All Students',
        'Grade 5 Math 4s - All Students',
        'Grade 6 Math 4s - All Students',
        'Grade 7 Math 4s - All Students',
        'Grade 8 Math 4s - All Students']] .mean(axis=1)

dfy['NativeMath4'] = dfy[['Grade 3 Math 4s - American Indian or Alaska Native',
        'Grade 4 Math 4s - American Indian or Alaska Native',
        'Grade 5 Math 4s - American Indian or Alaska Native',
        'Grade 6 Math 4s - American Indian or Alaska Native',
        'Grade 7 Math 4s - American Indian or Alaska Native',
        'Grade 8 Math 4s - American Indian or Alaska Native']] .mean(axis=1)

dfy['AfricanAmericanMath4'] = dfy[['Grade 3 Math 4s - Black or African American',
        'Grade 4 Math 4s - Black or African American',
        'Grade 5 Math 4s - Black or African American',
        'Grade 6 Math 4s - Black or African American',
        'Grade 7 Math 4s - Black or African American',
        'Grade 8 Math 4s - Black or African American']] .mean(axis=1)

dfy['LatinoMath4'] = dfy[['Grade 3 Math 4s - Hispanic or Latino',
        'Grade 4 Math 4s - Hispanic or Latino',
        'Grade 5 Math 4s - Hispanic or Latino',
        'Grade 6 Math 4s - Hispanic or Latino',
        'Grade 7 Math 4s - Hispanic or Latino',
        'Grade 8 Math 4s - Hispanic or Latino']] .mean(axis=1)

dfy['IslanderMath4'] = dfy[['Grade 3 Math 4s - Asian or Pacific Islander',
        'Grade 4 Math 4s - Asian or Pacific Islander',
        'Grade 5 Math 4s - Asian or Pacific Islander',
        'Grade 6 Math 4s - Asian or Pacific Islander',
        'Grade 7 Math 4s - Asian or Pacific Islander',
        'Grade 8 Math 4s - Asian or Pacific Islander']] .mean(axis=1)

dfy['WhiteMath4'] = dfy[['Grade 3 Math 4s - White',
        'Grade 4 Math 4s - White',
        'Grade 5 Math 4s - White',
        'Grade 6 Math 4s - White',
        'Grade 7 Math 4s - White',
        'Grade 8 Math 4s - White']] .mean(axis=1)

dfy['MultiracialMath4'] = dfy[['Grade 3 ELA 4s - Multiracial',

```

```

        'Grade 4 Math 4s - Multiracial',
        'Grade 5 Math 4s - Multiracial',
        'Grade 6 Math 4s - Multiracial',
        'Grade 7 Math 4s - Multiracial',
        'Grade 8 Math 4s - Multiracial']] .mean(axis=1)

dfy['LimitedEnglishMath4'] = dfy[['Grade 3 Math 4s - Limited English Proficient',
        'Grade 4 Math 4s - Limited English Proficient',
        'Grade 5 Math 4s - Limited English Proficient',
        'Grade 6 Math 4s - Limited English Proficient',
        'Grade 7 Math 4s - Limited English Proficient',
        'Grade 8 Math 4s - Limited English Proficient']] .mean(axis=1)

dfy['DisadvMath4'] = dfy[['Grade 3 Math 4s - Economically Disadvantaged',
        'Grade 4 Math 4s - Economically Disadvantaged',
        'Grade 5 Math 4s - Economically Disadvantaged',
        'Grade 6 Math 4s - Economically Disadvantaged',
        'Grade 7 Math 4s - Economically Disadvantaged',
        'Grade 8 Math 4s - Economically Disadvantaged']] .mean(axis=1)

In [588]: # drop columns
dfy= dfy.drop(['Grade 3 ELA - All Students Tested',
        'Grade 4 ELA - All Students Tested',
        'Grade 5 ELA - All Students Tested',
        'Grade 6 ELA - All Students Tested',
        'Grade 7 ELA - All Students Tested',
        'Grade 8 ELA - All Students Tested',
        'Grade 3 ELA 4s - All Students',
        'Grade 4 ELA 4s - All Students',
        'Grade 5 ELA 4s - All Students',
        'Grade 6 ELA 4s - All Students',
        'Grade 7 ELA 4s - All Students',
        'Grade 8 ELA 4s - All Students',
        'Grade 3 ELA 4s - American Indian or Alaska Native',
        'Grade 4 ELA 4s - American Indian or Alaska Native',
        'Grade 5 ELA 4s - American Indian or Alaska Native',
        'Grade 6 ELA 4s - American Indian or Alaska Native',
        'Grade 7 ELA 4s - American Indian or Alaska Native',
        'Grade 8 ELA 4s - American Indian or Alaska Native',
        'Grade 3 ELA 4s - Black or African American',
        'Grade 4 ELA 4s - Black or African American',
        'Grade 5 ELA 4s - Black or African American',
        'Grade 6 ELA 4s - Black or African American',
        'Grade 7 ELA 4s - Black or African American',
        'Grade 8 ELA 4s - Black or African American',
        'Grade 3 ELA 4s - Hispanic or Latino',
        'Grade 4 ELA 4s - Hispanic or Latino',
        'Grade 5 ELA 4s - Hispanic or Latino',

```

'Grade 6 ELA 4s - Hispanic or Latino',
 'Grade 7 ELA 4s - Hispanic or Latino',
 'Grade 8 ELA 4s - Hispanic or Latino',
 'Grade 3 ELA 4s - Asian or Pacific Islander',
 'Grade 4 ELA 4s - Asian or Pacific Islander',
 'Grade 5 ELA 4s - Asian or Pacific Islander',
 'Grade 6 ELA 4s - Asian or Pacific Islander',
 'Grade 7 ELA 4s - Asian or Pacific Islander',
 'Grade 8 ELA 4s - Asian or Pacific Islander',
 'Grade 3 ELA 4s - White',
 'Grade 4 ELA 4s - White',
 'Grade 5 ELA 4s - White',
 'Grade 6 ELA 4s - White',
 'Grade 7 ELA 4s - White',
 'Grade 8 ELA 4s - White',
 'Grade 3 ELA 4s - Multiracial',
 'Grade 4 ELA 4s - Multiracial',
 'Grade 5 ELA 4s - Multiracial',
 'Grade 6 ELA 4s - Multiracial',
 'Grade 7 ELA 4s - Multiracial',
 'Grade 8 ELA 4s - Multiracial',
 'Grade 3 ELA 4s - Limited English Proficient',
 'Grade 4 ELA 4s - Limited English Proficient',
 'Grade 5 ELA 4s - Limited English Proficient',
 'Grade 6 ELA 4s - Limited English Proficient',
 'Grade 7 ELA 4s - Limited English Proficient',
 'Grade 8 ELA 4s - Limited English Proficient',
 'Grade 3 ELA 4s - Economically Disadvantaged',
 'Grade 4 ELA 4s - Economically Disadvantaged',
 'Grade 5 ELA 4s - Economically Disadvantaged',
 'Grade 6 ELA 4s - Economically Disadvantaged',
 'Grade 7 ELA 4s - Economically Disadvantaged',
 'Grade 8 ELA 4s - Economically Disadvantaged',
 'Grade 3 Math - All Students tested',
 'Grade 4 Math - All Students Tested',
 'Grade 5 Math - All Students Tested',
 'Grade 6 Math - All Students Tested',
 'Grade 7 Math - All Students Tested',
 'Grade 8 Math - All Students Tested',
 'Grade 3 Math 4s - All Students',
 'Grade 4 Math 4s - All Students',
 'Grade 5 Math 4s - All Students',
 'Grade 6 Math 4s - All Students',
 'Grade 7 Math 4s - All Students',
 'Grade 8 Math 4s - All Students',
 'Grade 3 Math 4s - American Indian or Alaska Native',
 'Grade 4 Math 4s - American Indian or Alaska Native',
 'Grade 5 Math 4s - American Indian or Alaska Native',


```

        'Grade 6 Math 4s - American Indian or Alaska Native',
        'Grade 7 Math 4s - American Indian or Alaska Native',
        'Grade 8 Math 4s - American Indian or Alaska Native',
'Grade 3 Math 4s - Black or African American',
        'Grade 4 Math 4s - Black or African American',
        'Grade 5 Math 4s - Black or African American',
        'Grade 6 Math 4s - Black or African American',
        'Grade 7 Math 4s - Black or African American',
        'Grade 8 Math 4s - Black or African American',
'Grade 3 Math 4s - Hispanic or Latino',
        'Grade 4 Math 4s - Hispanic or Latino',
        'Grade 5 Math 4s - Hispanic or Latino',
        'Grade 6 Math 4s - Hispanic or Latino',
        'Grade 7 Math 4s - Hispanic or Latino',
        'Grade 8 Math 4s - Hispanic or Latino',
'Grade 3 Math 4s - Asian or Pacific Islander',
        'Grade 4 Math 4s - Asian or Pacific Islander',
        'Grade 5 Math 4s - Asian or Pacific Islander',
        'Grade 6 Math 4s - Asian or Pacific Islander',
        'Grade 7 Math 4s - Asian or Pacific Islander',
        'Grade 8 Math 4s - Asian or Pacific Islander',
'Grade 3 Math 4s - White',
        'Grade 4 Math 4s - White',
        'Grade 5 Math 4s - White',
        'Grade 6 Math 4s - White',
        'Grade 7 Math 4s - White',
        'Grade 8 Math 4s - White',
'Grade 3 ELA 4s - Multiracial',
        'Grade 4 Math 4s - Multiracial',
        'Grade 5 Math 4s - Multiracial',
        'Grade 6 Math 4s - Multiracial',
        'Grade 7 Math 4s - Multiracial',
        'Grade 8 Math 4s - Multiracial',
'Grade 3 Math 4s - Limited English Proficient',
        'Grade 4 Math 4s - Limited English Proficient',
        'Grade 5 Math 4s - Limited English Proficient',
        'Grade 6 Math 4s - Limited English Proficient',
        'Grade 7 Math 4s - Limited English Proficient',
        'Grade 8 Math 4s - Limited English Proficient',
'Grade 3 Math 4s - Economically Disadvantaged',
        'Grade 4 Math 4s - Economically Disadvantaged',
        'Grade 5 Math 4s - Economically Disadvantaged',
        'Grade 6 Math 4s - Economically Disadvantaged',
        'Grade 7 Math 4s - Economically Disadvantaged',
        'Grade 8 Math 4s - Economically Disadvantaged',
'Adjusted Grade',
'New?','Other Location Code in LCGMS'], axis=1)

```

```
In [589]: dfy.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1364 entries, 0 to 1363
Data columns (total 66 columns):
School Name                1364 non-null object
SED Code                   1364 non-null int64
Location Code              1364 non-null object
District                  1364 non-null int64
Latitude                  1364 non-null float64
Longitude                 1364 non-null float64
Address (Full)            1364 non-null object
City                      1364 non-null object
Zip                       1364 non-null int64
Grades                    1364 non-null object
Grade Low                 1364 non-null object
Grade High                1364 non-null object
Community School?         1364 non-null object
Economic Need Index       1339 non-null float64
School Income Estimate    906 non-null object
Percent ELL               1364 non-null object
Percent Asian             1364 non-null object
Percent Black             1364 non-null object
Percent Hispanic          1364 non-null object
Percent Black / Hispanic  1364 non-null object
Percent White             1364 non-null object
Student Attendance Rate   1339 non-null object
Percent of Students Chronically Absent 1339 non-null object
Rigorous Instruction %    1339 non-null object
Rigorous Instruction Rating 1288 non-null object
Collaborative Teachers %  1339 non-null object
Collaborative Teachers Rating 1288 non-null object
Supportive Environment %  1339 non-null object
Supportive Environment Rating 1284 non-null object
Effective School Leadership % 1339 non-null object
Effective School Leadership Rating 1291 non-null object
Strong Family-Community Ties % 1339 non-null object
Strong Family-Community Ties Rating 1291 non-null object
Trust %                   1339 non-null object
Trust Rating              1291 non-null object
Student Achievement Rating 1278 non-null object
Average ELA Proficiency   1309 non-null float64
Average Math Proficiency  1309 non-null float64
Grade 3 Math 4s - Multiracial 1364 non-null int64
DBN                       113 non-null object
School name               113 non-null object
Year of SHST              113 non-null float64
Grade level               113 non-null float64
```

```

Enrollment on 10/31                113 non-null float64
Number of students who registered for the SHSAT  113 non-null float64
Number of students who took the SHSAT          113 non-null float64
AllTested                                1364 non-null float64
All4                                      1364 non-null float64
Native4                                  1364 non-null float64
AfricanAmerican4                       1364 non-null float64
Latino4                                 1364 non-null float64
Islander4                              1364 non-null float64
White4                                 1364 non-null float64
Multiracial4                           1364 non-null float64
LimitedEnglish4                        1364 non-null float64
Disadv4                                1364 non-null float64
AllMath4Tested                         1364 non-null float64
AllMath4                               1364 non-null float64
NativeMath4                            1364 non-null float64
AfricanAmericanMath4                  1364 non-null float64
LatinoMath4                           1364 non-null float64
IslanderMath4                         1364 non-null float64
WhiteMath4                            1364 non-null float64
MultiracialMath4                      1364 non-null float64
LimitedEnglishMath4                  1364 non-null float64
DisadvMath4                           1364 non-null float64
dtypes: float64(30), int64(4), object(32)
memory usage: 714.0+ KB

```

```
In [590]: dfy.describe()
```

```

Out[590]:
      count  SED Code  District  Latitude  Longitude  Zip \
count  1.364000e+03  1364.000000  1364.000000  1364.000000  1364.000000
mean    3.274443e+11  15.384164   40.739964  -73.920300  10762.730938
std     1.265045e+10   9.354700   0.086044   0.078165   548.074632
min     3.075000e+11   1.000000   40.507803  -74.244025  10001.000000
25%     3.207000e+11   7.000000   40.672706  -73.955847  10304.000000
50%     3.314000e+11  14.000000   40.736511  -73.925881  10738.000000
75%     3.332000e+11  24.000000   40.816396  -73.884761  11228.000000
max     3.531009e+11  32.000000   40.903455  -73.708920  11694.000000

      Economic Need Index  Average ELA Proficiency  Average Math Proficiency \
count          1339.000000          1309.000000          1309.000000
mean             0.677808             2.530527             2.667540
std             0.205944             0.361546             0.473053
min             0.049000             1.810000             1.830000
25%            0.559500             2.260000             2.290000
50%            0.737000             2.450000             2.580000
75%            0.839000             2.740000             2.990000
max            0.957000             3.930000             4.200000

```

	Grade 3 Math 4s - Multiracial	Year of SHST	...	\
count	1364.000000	113.000000	...	
mean	0.061584	2014.601770	...	
std	0.538215	1.122337	...	
min	0.000000	2013.000000	...	
25%	0.000000	2014.000000	...	
50%	0.000000	2015.000000	...	
75%	0.000000	2016.000000	...	
max	8.000000	2016.000000	...	

	AllMath4Tested	AllMath4	NativeMath4	AfricanAmericanMath4	\
count	1364.000000	1364.000000	1364.000000	1364.000000	
mean	55.016984	10.307674	0.021505	1.526393	
std	41.252793	15.106568	0.179063	4.058560	
min	0.000000	0.000000	0.000000	0.000000	
25%	29.500000	1.333333	0.000000	0.000000	
50%	46.166667	4.333333	0.000000	0.166667	
75%	67.208333	13.833333	0.000000	1.166667	
max	330.333333	151.666667	3.500000	61.166667	

	LatinoMath4	IslanderMath4	WhiteMath4	MultiracialMath4	\
count	1364.000000	1364.000000	1364.000000	1364.000000	
mean	2.028715	3.249756	2.288368	0.046676	
std	3.480549	8.994379	6.348200	0.292881	
min	0.000000	0.000000	0.000000	0.000000	
25%	0.166667	0.000000	0.000000	0.000000	
50%	0.833333	0.000000	0.000000	0.000000	
75%	2.500000	1.500000	0.833333	0.000000	
max	29.666667	107.666667	88.000000	6.666667	

	LimitedEnglishMath4	DisadvMath4
count	1364.000000	1364.000000
mean	0.262830	5.353372
std	0.912010	8.654386
min	0.000000	0.000000
25%	0.000000	0.666667
50%	0.000000	2.166667
75%	0.166667	6.375000
max	14.333333	88.166667

[8 rows x 34 columns]

```
In [591]: dfy = dfy.merge(df3, how='left', right_on='Feeder School DBN' , left_on='Location Co
dfy.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1364 entries, 0 to 1363
```

Data columns (total 71 columns):

School Name	1364 non-null object
SED Code	1364 non-null int64
Location Code	1364 non-null object
District	1364 non-null int64
Latitude	1364 non-null float64
Longitude	1364 non-null float64
Address (Full)	1364 non-null object
City	1364 non-null object
Zip	1364 non-null int64
Grades	1364 non-null object
Grade Low	1364 non-null object
Grade High	1364 non-null object
Community School?	1364 non-null object
Economic Need Index	1339 non-null float64
School Income Estimate	906 non-null object
Percent ELL	1364 non-null object
Percent Asian	1364 non-null object
Percent Black	1364 non-null object
Percent Hispanic	1364 non-null object
Percent Black / Hispanic	1364 non-null object
Percent White	1364 non-null object
Student Attendance Rate	1339 non-null object
Percent of Students Chronically Absent	1339 non-null object
Rigorous Instruction %	1339 non-null object
Rigorous Instruction Rating	1288 non-null object
Collaborative Teachers %	1339 non-null object
Collaborative Teachers Rating	1288 non-null object
Supportive Environment %	1339 non-null object
Supportive Environment Rating	1284 non-null object
Effective School Leadership %	1339 non-null object
Effective School Leadership Rating	1291 non-null object
Strong Family-Community Ties %	1339 non-null object
Strong Family-Community Ties Rating	1291 non-null object
Trust %	1339 non-null object
Trust Rating	1291 non-null object
Student Achievement Rating	1278 non-null object
Average ELA Proficiency	1309 non-null float64
Average Math Proficiency	1309 non-null float64
Grade 3 Math 4s - Multiracial	1364 non-null int64
DBN	113 non-null object
School name	113 non-null object
Year of SHST	113 non-null float64
Grade level	113 non-null float64
Enrollment on 10/31	113 non-null float64
Number of students who registered for the SHSAT	113 non-null float64
Number of students who took the SHSAT	113 non-null float64
AllTested	1364 non-null float64

All4	1364 non-null float64
Native4	1364 non-null float64
AfricanAmerican4	1364 non-null float64
Latino4	1364 non-null float64
Islander4	1364 non-null float64
White4	1364 non-null float64
Multiracial4	1364 non-null float64
LimitedEnglish4	1364 non-null float64
Disadv4	1364 non-null float64
AllMath4Tested	1364 non-null float64
AllMath4	1364 non-null float64
NativeMath4	1364 non-null float64
AfricanAmericanMath4	1364 non-null float64
LatinoMath4	1364 non-null float64
IslanderMath4	1364 non-null float64
WhiteMath4	1364 non-null float64
MultiracialMath4	1364 non-null float64
LimitedEnglishMath4	1364 non-null float64
DisadvMath4	1364 non-null float64
Feeder School DBN	683 non-null object
Feeder School Name	683 non-null object
Count of Students in HS Admissions	683 non-null float64
Count of Testers	683 non-null float64
Count of Offers	683 non-null float64

dtypes: float64(33), int64(4), object(34)
memory usage: 767.2+ KB

In [592]: list(dfy)

Out[592]: ['School Name',
 'SED Code',
 'Location Code',
 'District',
 'Latitude',
 'Longitude',
 'Address (Full)',
 'City',
 'Zip',
 'Grades',
 'Grade Low',
 'Grade High',
 'Community School?',
 'Economic Need Index',
 'School Income Estimate',
 'Percent ELL',
 'Percent Asian',
 'Percent Black',

'Percent Hispanic',
 'Percent Black / Hispanic',
 'Percent White',
 'Student Attendance Rate',
 'Percent of Students Chronically Absent',
 'Rigorous Instruction %',
 'Rigorous Instruction Rating',
 'Collaborative Teachers %',
 'Collaborative Teachers Rating',
 'Supportive Environment %',
 'Supportive Environment Rating',
 'Effective School Leadership %',
 'Effective School Leadership Rating',
 'Strong Family-Community Ties %',
 'Strong Family-Community Ties Rating',
 'Trust %',
 'Trust Rating',
 'Student Achievement Rating',
 'Average ELA Proficiency',
 'Average Math Proficiency',
 'Grade 3 Math 4s - Multiracial',
 'DBN',
 'School name',
 'Year of SHST',
 'Grade level',
 'Enrollment on 10/31',
 'Number of students who registered for the SHSAT',
 'Number of students who took the SHSAT',
 'AllTested',
 'All4',
 'Native4',
 'AfricanAmerican4',
 'Latino4',
 'Islander4',
 'White4',
 'Multiracial4',
 'LimitedEnglish4',
 'Disadv4',
 'AllMath4Tested',
 'AllMath4',
 'NativeMath4',
 'AfricanAmericanMath4',
 'LatinoMath4',
 'IslanderMath4',
 'WhiteMath4',
 'MultiracialMath4',
 'LimitedEnglishMath4',
 'DisadvMath4',

```

'Feeder School DBN',
'Feeder School Name',
'Count of Students in HS Admissions',
'Count of Testers',
'Count of Offers']

```

In [593]: dfy.head(3)

```

Out[593]:
      School Name      SED Code Location Code  District  Latitude \
0  P.S. 015 ROBERTO CLEMENTE  310100010015      01M015         1  40.721834
1      P.S. 019 ASHER LEVY  310100010019      01M019         1  40.729892
2      P.S. 020 ANNA SILVER  310100010020      01M020         1  40.721274

      Longitude      Address (Full)      City      Zip \
0 -73.978766  333 E 4TH ST NEW YORK, NY 10009  NEW YORK  10009
1 -73.984231  185 1ST AVE NEW YORK, NY 10003  NEW YORK  10003
2 -73.986315  166 ESSEX ST NEW YORK, NY 10002  NEW YORK  10002

      Grades      ...      IslanderMath4 WhiteMath4 \
0  PK,OK,01,02,03,04,05  ...      0.000000  0.000000
1  PK,OK,01,02,03,04,05  ...      0.333333  0.000000
2  PK,OK,01,02,03,04,05  ...      4.500000  0.333333

      MultiracialMath4  LimitedEnglishMath4  DisadvMath4  Feeder School DBN \
0          0.0          0.000000      0.000000          NaN
1          0.0          0.000000      2.500000          NaN
2          0.0          0.166667      2.666667          NaN

      Feeder School Name  Count of Students in HS Admissions  Count of Testers \
0          NaN          NaN          NaN
1          NaN          NaN          NaN
2          NaN          NaN          NaN

      Count of Offers
0          NaN
1          NaN
2          NaN

[3 rows x 71 columns]

```

In [594]: dfy.corr()

```

Out[594]:
      SED Code      District  Latitude \
SED Code      1.000000  0.954898 -0.661821
District      0.954898  1.000000 -0.642761
Latitude     -0.661821 -0.642761  1.000000
Longitude      0.118464  0.178117  0.291984
Zip           0.775995  0.772984 -0.604029
Economic Need Index -0.311408 -0.283162  0.306643

```


Average ELA Proficiency	0.128419	0.111628	-0.195998
Average Math Proficiency	0.117325	0.108624	-0.166521
Grade 3 Math 4s - Multiracial	-0.039204	-0.054976	-0.026301
Year of SHST	0.033258	NaN	-0.052813
Grade level	0.154903	NaN	-0.159922
Enrollment on 10/31	-0.144704	NaN	0.339524
Number of students who registered for the SHSAT	0.011083	NaN	0.032252
Number of students who took the SHSAT	0.007385	NaN	-0.121631
AllTested	0.238549	0.248214	-0.106065
All4	0.181169	0.176803	-0.183555
Native4	0.109187	0.113853	-0.056912
AfricanAmerican4	-0.008202	0.006479	-0.069582
Latino4	0.084169	0.090725	0.091737
Islander4	0.217653	0.220223	-0.129859
White4	0.109849	0.090336	-0.233920
Multiracial4	-0.101175	-0.103340	-0.021333
LimitedEnglish4	0.063333	0.051208	-0.025918
Disadv4	0.222882	0.238323	-0.154559
AllMath4Tested	0.237845	0.249322	-0.100759
AllMath4	0.144911	0.145041	-0.162547
NativeMath4	0.103002	0.107702	-0.049700
AfricanAmericanMath4	-0.121024	-0.104757	0.030972
LatinoMath4	0.005167	0.016430	0.130747
IslanderMath4	0.222771	0.223257	-0.153146
WhiteMath4	0.111477	0.094061	-0.248455
MultiracialMath4	-0.100134	-0.097328	-0.023081
LimitedEnglishMath4	0.101698	0.097299	-0.099548
DisadvMath4	0.172113	0.189524	-0.126465
Count of Students in HS Admissions	0.355074	0.331419	-0.202395
Count of Testers	0.281210	0.260098	-0.220273
Count of Offers	0.119194	0.109803	-0.145873

	Longitude	Zip \
SED Code	0.118464	0.775995
District	0.178117	0.772984
Latitude	0.291984	-0.604029
Longitude	1.000000	0.411442
Zip	0.411442	1.000000
Economic Need Index	0.015074	-0.165680
Average ELA Proficiency	-0.088961	0.059880
Average Math Proficiency	-0.080416	0.060285
Grade 3 Math 4s - Multiracial	-0.095886	-0.026360
Year of SHST	0.020974	-0.027054
Grade level	0.139795	0.054755
Enrollment on 10/31	0.233728	0.367929
Number of students who registered for the SHSAT	-0.399194	-0.209484
Number of students who took the SHSAT	-0.259444	-0.148144
AllTested	0.023724	0.160152

All4	-0.055796	0.129721
Native4	0.116855	0.111561
AfricanAmerican4	0.069077	0.063760
Latino4	0.019831	0.039055
Islander4	0.104443	0.211271
White4	-0.245419	0.002639
Multiracial4	-0.099501	-0.082597
LimitedEnglish4	0.002595	0.086197
Disadv4	0.046138	0.208663
AllMath4Tested	0.023147	0.162333
AllMath4	-0.063211	0.115274
NativeMath4	0.119862	0.103137
AfricanAmericanMath4	0.009498	-0.064445
LatinoMath4	0.001571	-0.021385
IslanderMath4	0.084367	0.231808
WhiteMath4	-0.251764	0.010609
MultiracialMath4	-0.103052	-0.079925
LimitedEnglishMath4	-0.011005	0.137108
DisadvMath4	0.025356	0.178356
Count of Students in HS Admissions	0.001274	0.241836
Count of Testers	-0.019732	0.231809
Count of Offers	-0.048907	0.107895

	Economic Need Index \
SED Code	-0.311408
District	-0.283162
Latitude	0.306643
Longitude	0.015074
Zip	-0.165680
Economic Need Index	1.000000
Average ELA Proficiency	-0.800394
Average Math Proficiency	-0.702374
Grade 3 Math 4s - Multiracial	-0.240659
Year of SHST	-0.011539
Grade level	-0.332009
Enrollment on 10/31	-0.192580
Number of students who registered for the SHSAT	0.023440
Number of students who took the SHSAT	-0.172916
AllTested	-0.180824
All4	-0.510312
Native4	-0.062567
AfricanAmerican4	-0.072707
Latino4	-0.127446
Islander4	-0.323675
White4	-0.549216
Multiracial4	-0.281811
LimitedEnglish4	0.034956
Disadv4	-0.270175

AllMath4Tested	-0.155151
AllMath4	-0.450023
NativeMath4	-0.059500
AfricanAmericanMath4	0.004962
LatinoMath4	-0.029510
IslanderMath4	-0.286434
WhiteMath4	-0.566390
MultiracialMath4	-0.266687
LimitedEnglishMath4	-0.015378
DisadvMath4	-0.209139
Count of Students in HS Admissions	-0.252197
Count of Testers	-0.411867
Count of Offers	-0.421596

	Average ELA Proficiency \
SED Code	0.128419
District	0.111628
Latitude	-0.195998
Longitude	-0.088961
Zip	0.059880
Economic Need Index	-0.800394
Average ELA Proficiency	1.000000
Average Math Proficiency	0.929950
Grade 3 Math 4s - Multiracial	0.196213
Year of SHST	0.012700
Grade level	0.243363
Enrollment on 10/31	0.094302
Number of students who registered for the SHSAT	0.111542
Number of students who took the SHSAT	0.283662
AllTested	0.236735
All4	0.658177
Native4	0.054658
AfricanAmerican4	0.212182
Latino4	0.326014
Islander4	0.440704
White4	0.566242
Multiracial4	0.338812
LimitedEnglish4	0.110526
Disadv4	0.462595
AllMath4Tested	0.206157
AllMath4	0.637720
NativeMath4	0.060073
AfricanAmericanMath4	0.161203
LatinoMath4	0.240882
IslanderMath4	0.407912
WhiteMath4	0.568440
MultiracialMath4	0.327528
LimitedEnglishMath4	0.124413

DisadvMath4	0.427474
Count of Students in HS Admissions	0.210394
Count of Testers	0.465485
Count of Offers	0.520695

	Average Math Proficiency \
SED Code	0.117325
District	0.108624
Latitude	-0.166521
Longitude	-0.080416
Zip	0.060285
Economic Need Index	-0.702374
Average ELA Proficiency	0.929950
Average Math Proficiency	1.000000
Grade 3 Math 4s - Multiracial	0.175799
Year of SHST	0.007885
Grade level	0.265510
Enrollment on 10/31	-0.013963
Number of students who registered for the SHSAT	0.078433
Number of students who took the SHSAT	0.223685
AllTested	0.251121
All4	0.614635
Native4	0.045762
AfricanAmerican4	0.225106
Latino4	0.328540
Islander4	0.437103
White4	0.483199
Multiracial4	0.286675
LimitedEnglish4	0.171579
Disadv4	0.466593
AllMath4Tested	0.232278
AllMath4	0.681759
NativeMath4	0.065718
AfricanAmericanMath4	0.263405
LatinoMath4	0.326734
IslanderMath4	0.442585
WhiteMath4	0.501042
MultiracialMath4	0.284416
LimitedEnglishMath4	0.224012
DisadvMath4	0.517739
Count of Students in HS Admissions	0.194387
Count of Testers	0.442862
Count of Offers	0.468433

	Grade 3 Math 4s - Multiracial \
SED Code	-0.039204
District	-0.054976
Latitude	-0.026301

Longitude	-0.095886
Zip	-0.026360
Economic Need Index	-0.240659
Average ELA Proficiency	0.196213
Average Math Proficiency	0.175799
Grade 3 Math 4s - Multiracial	1.000000
Year of SHST	NaN
Grade level	NaN
Enrollment on 10/31	NaN
Number of students who registered for the SHSAT	NaN
Number of students who took the SHSAT	NaN
AllTested	0.022525
All4	0.129934
Native4	-0.012703
AfricanAmerican4	0.001654
Latino4	0.018822
Islander4	0.029978
White4	0.202957
Multiracial4	0.300033
LimitedEnglish4	-0.026729
Disadv4	0.008969
AllMath4Tested	0.023866
AllMath4	0.116946
NativeMath4	-0.013752
AfricanAmericanMath4	-0.011045
LatinoMath4	0.011719
IslanderMath4	0.016346
WhiteMath4	0.226853
MultiracialMath4	0.287386
LimitedEnglishMath4	-0.019547
DisadvMath4	-0.006277
Count of Students in HS Admissions	-0.023242
Count of Testers	0.022784
Count of Offers	0.030693

	Year of SHST \
SED Code	0.033258
District	NaN
Latitude	-0.052813
Longitude	0.020974
Zip	-0.027054
Economic Need Index	-0.011539
Average ELA Proficiency	0.012700
Average Math Proficiency	0.007885
Grade 3 Math 4s - Multiracial	NaN
Year of SHST	1.000000
Grade level	0.050325
Enrollment on 10/31	-0.061249

Number of students who registered for the SHSAT	-0.153356
Number of students who took the SHSAT	-0.045207
AllTested	-0.006235
All4	0.009377
Native4	NaN
AfricanAmerican4	0.083052
Latino4	-0.025164
Islander4	-0.025141
White4	-0.025141
Multiracial4	-0.025141
LimitedEnglish4	-0.021568
Disadv4	0.041013
AllMath4Tested	-0.009201
AllMath4	0.019439
NativeMath4	NaN
AfricanAmericanMath4	0.055281
LatinoMath4	-0.019891
IslanderMath4	-0.025141
WhiteMath4	-0.025141
MultiracialMath4	-0.010211
LimitedEnglishMath4	-0.013455
DisadvMath4	0.032070
Count of Students in HS Admissions	-0.015943
Count of Testers	0.017560
Count of Offers	-0.025141

SED Code	...
District	...
Latitude	...
Longitude	...
Zip	...
Economic Need Index	...
Average ELA Proficiency	...
Average Math Proficiency	...
Grade 3 Math 4s - Multiracial	...
Year of SHST	...
Grade level	...
Enrollment on 10/31	...
Number of students who registered for the SHSAT	...
Number of students who took the SHSAT	...
AllTested	...
All4	...
Native4	...
AfricanAmerican4	...
Latino4	...
Islander4	...
White4	...

\

Multiracial4	...
LimitedEnglish4	...
Disadv4	...
AllMath4Tested	...
AllMath4	...
NativeMath4	...
AfricanAmericanMath4	...
LatinoMath4	...
IslanderMath4	...
WhiteMath4	...
MultiracialMath4	...
LimitedEnglishMath4	...
DisadvMath4	...
Count of Students in HS Admissions	...
Count of Testers	...
Count of Offers	...

	AfricanAmericanMath4 \
SED Code	-0.121024
District	-0.104757
Latitude	0.030972
Longitude	0.009498
Zip	-0.064445
Economic Need Index	0.004962
Average ELA Proficiency	0.161203
Average Math Proficiency	0.263405
Grade 3 Math 4s - Multiracial	-0.011045
Year of SHST	0.055281
Grade level	0.157422
Enrollment on 10/31	-0.237079
Number of students who registered for the SHSAT	-0.152962
Number of students who took the SHSAT	-0.036843
AllTested	0.133670
All4	0.123543
Native4	0.019590
AfricanAmerican4	0.839716
Latino4	0.058003
Islander4	-0.062393
White4	-0.044113
Multiracial4	0.020160
LimitedEnglish4	-0.037597
Disadv4	0.191805
AllMath4Tested	0.131229
AllMath4	0.269587
NativeMath4	0.076112
AfricanAmericanMath4	1.000000
LatinoMath4	0.161130
IslanderMath4	-0.077146

WhiteMath4	-0.048821
MultiracialMath4	0.094581
LimitedEnglishMath4	-0.042476
DisadvMath4	0.341562
Count of Students in HS Admissions	-0.054994
Count of Testers	0.018657
Count of Offers	-0.044907

	LatinoMath4	IslanderMath4 \
SED Code	0.005167	0.222771
District	0.016430	0.223257
Latitude	0.130747	-0.153146
Longitude	0.001571	0.084367
Zip	-0.021385	0.231808
Economic Need Index	-0.029510	-0.286434
Average ELA Proficiency	0.240882	0.407912
Average Math Proficiency	0.326734	0.442585
Grade 3 Math 4s - Multiracial	0.011719	0.016346
Year of SHST	-0.019891	-0.025141
Grade level	0.033879	0.113581
Enrollment on 10/31	-0.055517	0.033151
Number of students who registered for the SHSAT	0.288104	0.101141
Number of students who took the SHSAT	0.080072	0.291501
AllTested	0.484628	0.571474
All4	0.356964	0.726540
Native4	0.004995	0.080127
AfricanAmerican4	0.098546	-0.054732
Latino4	0.833767	0.265917
Islander4	0.176991	0.940405
White4	0.136859	0.378943
Multiracial4	0.045885	0.146847
LimitedEnglish4	0.293614	0.225757
Disadv4	0.417239	0.761421
AllMath4Tested	0.493930	0.575876
AllMath4	0.463463	0.793577
NativeMath4	0.025953	0.080432
AfricanAmericanMath4	0.161130	-0.077146
LatinoMath4	1.000000	0.166465
IslanderMath4	0.166465	1.000000
WhiteMath4	0.132373	0.382159
MultiracialMath4	0.066978	0.145016
LimitedEnglishMath4	0.180779	0.610268
DisadvMath4	0.494340	0.788233
Count of Students in HS Admissions	0.352206	0.640299
Count of Testers	0.331389	0.790181
Count of Offers	0.157090	0.781842

WhiteMath4	MultiracialMath4 \
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SED Code	0.111477	-0.100134
District	0.094061	-0.097328
Latitude	-0.248455	-0.023081
Longitude	-0.251764	-0.103052
Zip	0.010609	-0.079925
Economic Need Index	-0.566390	-0.266687
Average ELA Proficiency	0.568440	0.327528
Average Math Proficiency	0.501042	0.284416
Grade 3 Math 4s - Multiracial	0.226853	0.287386
Year of SHST	-0.025141	-0.010211
Grade level	0.113581	0.135376
Enrollment on 10/31	0.033151	-0.003628
Number of students who registered for the SHSAT	0.101141	0.001986
Number of students who took the SHSAT	0.291501	0.167718
AllTested	0.402895	0.133575
All4	0.779631	0.404398
Native4	-0.011612	-0.010621
AfricanAmerican4	-0.004120	0.110634
Latino4	0.249102	0.121022
Islander4	0.443344	0.189189
White4	0.973047	0.488634
Multiracial4	0.447110	0.903375
LimitedEnglish4	0.009921	-0.034339
Disadv4	0.473337	0.135630
AllMath4Tested	0.386273	0.113615
AllMath4	0.700406	0.348437
NativeMath4	-0.010641	-0.012548
AfricanAmericanMath4	-0.048821	0.094581
LatinoMath4	0.132373	0.066978
IslanderMath4	0.382159	0.145016
WhiteMath4	1.000000	0.466884
MultiracialMath4	0.466884	1.000000
LimitedEnglishMath4	0.065524	-0.015748
DisadvMath4	0.388335	0.109534
Count of Students in HS Admissions	0.443029	0.130456
Count of Testers	0.643815	0.302159
Count of Offers	0.727092	0.532177

LimitedEnglishMath4 \

SED Code	0.101698
District	0.097299
Latitude	-0.099548
Longitude	-0.011005
Zip	0.137108
Economic Need Index	-0.015378
Average ELA Proficiency	0.124413
Average Math Proficiency	0.224012
Grade 3 Math 4s - Multiracial	-0.019547

Year of SHST	-0.013455
Grade level	-0.016051
Enrollment on 10/31	-0.027152
Number of students who registered for the SHSAT	0.221893
Number of students who took the SHSAT	0.013616
AllTested	0.358098
All4	0.254364
Native4	0.002829
AfricanAmerican4	-0.059808
Latino4	0.166169
Islander4	0.399347
White4	0.053820
Multiracial4	-0.019669
LimitedEnglish4	0.379997
Disadv4	0.328336
AllMath4Tested	0.383147
AllMath4	0.427443
NativeMath4	0.007294
AfricanAmericanMath4	-0.042476
LatinoMath4	0.180779
IslanderMath4	0.610268
WhiteMath4	0.065524
MultiracialMath4	-0.015748
LimitedEnglishMath4	1.000000
DisadvMath4	0.491741
Count of Students in HS Admissions	0.458174
Count of Testers	0.441553
Count of Offers	0.286620

	DisadvMath4 \
SED Code	0.172113
District	0.189524
Latitude	-0.126465
Longitude	0.025356
Zip	0.178356
Economic Need Index	-0.209139
Average ELA Proficiency	0.427474
Average Math Proficiency	0.517739
Grade 3 Math 4s - Multiracial	-0.006277
Year of SHST	0.032070
Grade level	0.111544
Enrollment on 10/31	-0.165340
Number of students who registered for the SHSAT	0.102049
Number of students who took the SHSAT	0.071933
AllTested	0.697246
All4	0.746844
Native4	0.086700
AfricanAmerican4	0.294488

Latino4	0.506948
Islander4	0.746271
White4	0.377947
Multiracial4	0.067370
LimitedEnglish4	0.233133
Disadv4	0.906080
AllMath4Tested	0.707533
AllMath4	0.879551
NativeMath4	0.110006
AfricanAmericanMath4	0.341562
LatinoMath4	0.494340
IslanderMath4	0.788233
WhiteMath4	0.388335
MultiracialMath4	0.109534
LimitedEnglishMath4	0.491741
DisadvMath4	1.000000
Count of Students in HS Admissions	0.611494
Count of Testers	0.732673
Count of Offers	0.599504

	Count of Students in HS Admissions
SED Code	0.355074
District	0.331419
Latitude	-0.202395
Longitude	0.001274
Zip	0.241836
Economic Need Index	-0.252197
Average ELA Proficiency	0.210394
Average Math Proficiency	0.194387
Grade 3 Math 4s - Multiracial	-0.023242
Year of SHST	-0.015943
Grade level	0.161501
Enrollment on 10/31	0.385689
Number of students who registered for the SHSAT	0.058574
Number of students who took the SHSAT	0.144917
AllTested	0.909962
All4	0.651056
Native4	0.204923
AfricanAmerican4	0.055765
Latino4	0.578781
Islander4	0.616307
White4	0.449475
Multiracial4	0.103813
LimitedEnglish4	0.230929
Disadv4	0.692586
AllMath4Tested	0.897209
AllMath4	0.612487
NativeMath4	0.198148

AfricanAmericanMath4	-0.054994
LatinoMath4	0.352206
IslanderMath4	0.640299
WhiteMath4	0.443029
MultiracialMath4	0.130456
LimitedEnglishMath4	0.458174
DisadvMath4	0.611494
Count of Students in HS Admissions	1.000000
Count of Testers	0.856769
Count of Offers	0.521510

	Count of Testers \
SED Code	0.281210
District	0.260098
Latitude	-0.220273
Longitude	-0.019732
Zip	0.231809
Economic Need Index	-0.411867
Average ELA Proficiency	0.465485
Average Math Proficiency	0.442862
Grade 3 Math 4s - Multiracial	0.022784
Year of SHST	0.017560
Grade level	0.159930
Enrollment on 10/31	0.111636
Number of students who registered for the SHSAT	0.109380
Number of students who took the SHSAT	0.333936
AllTested	0.804685
All4	0.852334
Native4	0.107553
AfricanAmerican4	0.134114
Latino4	0.541865
Islander4	0.780939
White4	0.645959
Multiracial4	0.281024
LimitedEnglish4	0.194009
Disadv4	0.807129
AllMath4Tested	0.782446
AllMath4	0.813443
NativeMath4	0.122513
AfricanAmericanMath4	0.018657
LatinoMath4	0.331389
IslanderMath4	0.790181
WhiteMath4	0.643815
MultiracialMath4	0.302159
LimitedEnglishMath4	0.441553
DisadvMath4	0.732673
Count of Students in HS Admissions	0.856769
Count of Testers	1.000000

Count of Offers	0.778770
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	Count of Offers
SED Code	0.119194
District	0.109803
Latitude	-0.145873
Longitude	-0.048907
Zip	0.107895
Economic Need Index	-0.421596
Average ELA Proficiency	0.520695
Average Math Proficiency	0.468433
Grade 3 Math 4s - Multiracial	0.030693
Year of SHST	-0.025141
Grade level	0.113581
Enrollment on 10/31	0.033151
Number of students who registered for the SHSAT	0.101141
Number of students who took the SHSAT	0.291501
AllTested	0.485583
All4	0.859680
Native4	0.038949
AfricanAmerican4	0.000243
Latino4	0.285010
Islander4	0.829191
White4	0.738594
Multiracial4	0.542744
LimitedEnglish4	0.103076
Disadv4	0.678692
AllMath4Tested	0.456146
AllMath4	0.792829
NativeMath4	0.061064
AfricanAmericanMath4	-0.044907
LatinoMath4	0.157090
IslanderMath4	0.781842
WhiteMath4	0.727092
MultiracialMath4	0.532177
LimitedEnglishMath4	0.286620
DisadvMath4	0.599504
Count of Students in HS Admissions	0.521510
Count of Testers	0.778770
Count of Offers	1.000000

[37 rows x 37 columns]

In this initial exploration, we can clearly observe that : Economic Need Index inversely correlated to Average ELA Proficiency and Economic Need Index inversely correlated to Average Math Proficiency

```
In [595]: df_sub=dfy
          df_sub=df_sub.drop(['District'],
```

```

'Latitude',
'Longitude',
'Address (Full)',
'City',
'Zip',
'Grades',
'Grade Low',
'Grade High',
'Rigorous Instruction %',
'Collaborative Teachers %',
'Supportive Environment %',
'Effective School Leadership %',
'Strong Family-Community Ties %',
'Trust %',
'Grade 3 Math 4s - Multiracial',
'DBN',
'School name',
'Year of SHST',
'Grade level',
'Enrollment on 10/31',
'Feeder School DBN',
'Feeder School Name'],axis=1)

```

```
In [596]: df_sub.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 1364 entries, 0 to 1363
Data columns (total 48 columns):
School Name                1364 non-null object
SED Code                   1364 non-null int64
Location Code              1364 non-null object
Community School?          1364 non-null object
Economic Need Index        1339 non-null float64
School Income Estimate      906 non-null object
Percent ELL                1364 non-null object
Percent Asian              1364 non-null object
Percent Black              1364 non-null object
Percent Hispanic           1364 non-null object
Percent Black / Hispanic   1364 non-null object
Percent White              1364 non-null object
Student Attendance Rate    1339 non-null object
Percent of Students Chronically Absent 1339 non-null object
Rigorous Instruction Rating 1288 non-null object
Collaborative Teachers Rating 1288 non-null object
Supportive Environment Rating 1284 non-null object
Effective School Leadership Rating 1291 non-null object
Strong Family-Community Ties Rating 1291 non-null object
Trust Rating               1291 non-null object

```

Student Achievement Rating	1278 non-null object
Average ELA Proficiency	1309 non-null float64
Average Math Proficiency	1309 non-null float64
Number of students who registered for the SHSAT	113 non-null float64
Number of students who took the SHSAT	113 non-null float64
AllTested	1364 non-null float64
All4	1364 non-null float64
Native4	1364 non-null float64
AfricanAmerican4	1364 non-null float64
Latino4	1364 non-null float64
Islander4	1364 non-null float64
White4	1364 non-null float64
Multiracial4	1364 non-null float64
LimitedEnglish4	1364 non-null float64
Disadv4	1364 non-null float64
AllMath4Tested	1364 non-null float64
AllMath4	1364 non-null float64
NativeMath4	1364 non-null float64
AfricanAmericanMath4	1364 non-null float64
LatinoMath4	1364 non-null float64
IslanderMath4	1364 non-null float64
WhiteMath4	1364 non-null float64
MultiracialMath4	1364 non-null float64
LimitedEnglishMath4	1364 non-null float64
DisadvMath4	1364 non-null float64
Count of Students in HS Admissions	683 non-null float64
Count of Testers	683 non-null float64
Count of Offers	683 non-null float64

dtypes: float64(28), int64(1), object(19)
memory usage: 522.2+ KB

In [597]: df_sub.corr()

Out [597]:	SED Code \
SED Code	1.000000
Economic Need Index	-0.311408
Average ELA Proficiency	0.128419
Average Math Proficiency	0.117325
Number of students who registered for the SHSAT	0.011083
Number of students who took the SHSAT	0.007385
AllTested	0.238549
All4	0.181169
Native4	0.109187
AfricanAmerican4	-0.008202
Latino4	0.084169
Islander4	0.217653
White4	0.109849

Multiracial4	-0.101175
LimitedEnglish4	0.063333
Disadv4	0.222882
AllMath4Tested	0.237845
AllMath4	0.144911
NativeMath4	0.103002
AfricanAmericanMath4	-0.121024
LatinoMath4	0.005167
IslanderMath4	0.222771
WhiteMath4	0.111477
MultiracialMath4	-0.100134
LimitedEnglishMath4	0.101698
DisadvMath4	0.172113
Count of Students in HS Admissions	0.355074
Count of Testers	0.281210
Count of Offers	0.119194

	Economic Need Index \
SED Code	-0.311408
Economic Need Index	1.000000
Average ELA Proficiency	-0.800394
Average Math Proficiency	-0.702374
Number of students who registered for the SHSAT	0.023440
Number of students who took the SHSAT	-0.172916
AllTested	-0.180824
All4	-0.510312
Native4	-0.062567
AfricanAmerican4	-0.072707
Latino4	-0.127446
Islander4	-0.323675
White4	-0.549216
Multiracial4	-0.281811
LimitedEnglish4	0.034956
Disadv4	-0.270175
AllMath4Tested	-0.155151
AllMath4	-0.450023
NativeMath4	-0.059500
AfricanAmericanMath4	0.004962
LatinoMath4	-0.029510
IslanderMath4	-0.286434
WhiteMath4	-0.566390
MultiracialMath4	-0.266687
LimitedEnglishMath4	-0.015378
DisadvMath4	-0.209139
Count of Students in HS Admissions	-0.252197
Count of Testers	-0.411867
Count of Offers	-0.421596

	Average ELA Proficiency \
SED Code	0.128419
Economic Need Index	-0.800394
Average ELA Proficiency	1.000000
Average Math Proficiency	0.929950
Number of students who registered for the SHSAT	0.111542
Number of students who took the SHSAT	0.283662
AllTested	0.236735
All4	0.658177
Native4	0.054658
AfricanAmerican4	0.212182
Latino4	0.326014
Islander4	0.440704
White4	0.566242
Multiracial4	0.338812
LimitedEnglish4	0.110526
Disadv4	0.462595
AllMath4Tested	0.206157
AllMath4	0.637720
NativeMath4	0.060073
AfricanAmericanMath4	0.161203
LatinoMath4	0.240882
IslanderMath4	0.407912
WhiteMath4	0.568440
MultiracialMath4	0.327528
LimitedEnglishMath4	0.124413
DisadvMath4	0.427474
Count of Students in HS Admissions	0.210394
Count of Testers	0.465485
Count of Offers	0.520695

	Average Math Proficiency \
SED Code	0.117325
Economic Need Index	-0.702374
Average ELA Proficiency	0.929950
Average Math Proficiency	1.000000
Number of students who registered for the SHSAT	0.078433
Number of students who took the SHSAT	0.223685
AllTested	0.251121
All4	0.614635
Native4	0.045762
AfricanAmerican4	0.225106
Latino4	0.328540
Islander4	0.437103
White4	0.483199
Multiracial4	0.286675
LimitedEnglish4	0.171579
Disadv4	0.466593

AllMath4Tested	0.232278
AllMath4	0.681759
NativeMath4	0.065718
AfricanAmericanMath4	0.263405
LatinoMath4	0.326734
IslanderMath4	0.442585
WhiteMath4	0.501042
MultiracialMath4	0.284416
LimitedEnglishMath4	0.224012
DisadvMath4	0.517739
Count of Students in HS Admissions	0.194387
Count of Testers	0.442862
Count of Offers	0.468433

Number of students who registered for

SED Code
Economic Need Index
Average ELA Proficiency
Average Math Proficiency
Number of students who registered for the SHSAT
Number of students who took the SHSAT
AllTested
All4
Native4
AfricanAmerican4
Latino4
Islander4
White4
Multiracial4
LimitedEnglish4
Disadv4
AllMath4Tested
AllMath4
NativeMath4
AfricanAmericanMath4
LatinoMath4
IslanderMath4
WhiteMath4
MultiracialMath4
LimitedEnglishMath4
DisadvMath4
Count of Students in HS Admissions
Count of Testers
Count of Offers

Number of students who took the SHSAT

SED Code	0.0073
Economic Need Index	-0.1729

Average ELA Proficiency	0.2836
Average Math Proficiency	0.2236
Number of students who registered for the SHSAT	0.7135
Number of students who took the SHSAT	1.0000
AllTested	-0.0603
All4	0.3166
Native4	NA
AfricanAmerican4	0.0989
Latino4	0.2543
Islander4	0.2915
White4	0.2915
Multiracial4	0.2915
LimitedEnglish4	-0.0139
Disadv4	0.3070
AllMath4Tested	-0.1005
AllMath4	0.1097
NativeMath4	NA
AfricanAmericanMath4	-0.0368
LatinoMath4	0.0800
IslanderMath4	0.2915
WhiteMath4	0.2915
MultiracialMath4	0.1677
LimitedEnglishMath4	0.0136
DisadvMath4	0.0719
Count of Students in HS Admissions	0.1449
Count of Testers	0.3339
Count of Offers	0.2915

	AllTested	All4 \
SED Code	0.238549	0.181169
Economic Need Index	-0.180824	-0.510312
Average ELA Proficiency	0.236735	0.658177
Average Math Proficiency	0.251121	0.614635
Number of students who registered for the SHSAT	-0.021819	0.200793
Number of students who took the SHSAT	-0.060369	0.316628
AllTested	1.000000	0.686314
All4	0.686314	1.000000
Native4	0.143783	0.089638
AfricanAmerican4	0.203450	0.217079
Latino4	0.643399	0.529362
Islander4	0.581499	0.813208
White4	0.421975	0.810412
Multiracial4	0.102096	0.389538
LimitedEnglish4	0.176506	0.128738
Disadv4	0.744634	0.856384
AllMath4Tested	0.993251	0.657475
AllMath4	0.681137	0.923672
NativeMath4	0.116180	0.085426

AfricanAmericanMath4	0.133670	0.123543
LatinoMath4	0.484628	0.356964
IslanderMath4	0.571474	0.726540
WhiteMath4	0.402895	0.779631
MultiracialMath4	0.133575	0.404398
LimitedEnglishMath4	0.358098	0.254364
DisadvMath4	0.697246	0.746844
Count of Students in HS Admissions	0.909962	0.651056
Count of Testers	0.804685	0.852334
Count of Offers	0.485583	0.859680

	Native4	AfricanAmerican4 \
SED Code	0.109187	-0.008202
Economic Need Index	-0.062567	-0.072707
Average ELA Proficiency	0.054658	0.212182
Average Math Proficiency	0.045762	0.225106
Number of students who registered for the SHSAT	NaN	-0.078216
Number of students who took the SHSAT	NaN	0.098985
AllTested	0.143783	0.203450
All4	0.089638	0.217079
Native4	1.000000	0.108634
AfricanAmerican4	0.108634	1.000000
Latino4	0.035452	0.095225
Islander4	0.113690	-0.019605
White4	-0.013632	0.012959
Multiracial4	-0.000053	0.050383
LimitedEnglish4	-0.025916	-0.054510
Disadv4	0.146602	0.284380
AllMath4Tested	0.131057	0.184188
AllMath4	0.058387	0.240237
NativeMath4	0.895855	0.110782
AfricanAmericanMath4	0.019590	0.839716
LatinoMath4	0.004995	0.098546
IslanderMath4	0.080127	-0.054732
WhiteMath4	-0.011612	-0.004120
MultiracialMath4	-0.010621	0.110634
LimitedEnglishMath4	0.002829	-0.059808
DisadvMath4	0.086700	0.294488
Count of Students in HS Admissions	0.204923	0.055765
Count of Testers	0.107553	0.134114
Count of Offers	0.038949	0.000243

	...	\
SED Code	...	
Economic Need Index	...	
Average ELA Proficiency	...	
Average Math Proficiency	...	
Number of students who registered for the SHSAT	...	

Number of students who took the SHSAT	...
AllTested	...
All4	...
Native4	...
AfricanAmerican4	...
Latino4	...
Islander4	...
White4	...
Multiracial4	...
LimitedEnglish4	...
Disadv4	...
AllMath4Tested	...
AllMath4	...
NativeMath4	...
AfricanAmericanMath4	...
LatinoMath4	...
IslanderMath4	...
WhiteMath4	...
MultiracialMath4	...
LimitedEnglishMath4	...
DisadvMath4	...
Count of Students in HS Admissions	...
Count of Testers	...
Count of Offers	...

	AfricanAmericanMath4 \
SED Code	-0.121024
Economic Need Index	0.004962
Average ELA Proficiency	0.161203
Average Math Proficiency	0.263405
Number of students who registered for the SHSAT	-0.152962
Number of students who took the SHSAT	-0.036843
AllTested	0.133670
All4	0.123543
Native4	0.019590
AfricanAmerican4	0.839716
Latino4	0.058003
Islander4	-0.062393
White4	-0.044113
Multiracial4	0.020160
LimitedEnglish4	-0.037597
Disadv4	0.191805
AllMath4Tested	0.131229
AllMath4	0.269587
NativeMath4	0.076112
AfricanAmericanMath4	1.000000
LatinoMath4	0.161130
IslanderMath4	-0.077146

WhiteMath4	-0.048821
MultiracialMath4	0.094581
LimitedEnglishMath4	-0.042476
DisadvMath4	0.341562
Count of Students in HS Admissions	-0.054994
Count of Testers	0.018657
Count of Offers	-0.044907

	LatinoMath4	IslanderMath4 \
SED Code	0.005167	0.222771
Economic Need Index	-0.029510	-0.286434
Average ELA Proficiency	0.240882	0.407912
Average Math Proficiency	0.326734	0.442585
Number of students who registered for the SHSAT	0.288104	0.101141
Number of students who took the SHSAT	0.080072	0.291501
AllTested	0.484628	0.571474
All4	0.356964	0.726540
Native4	0.004995	0.080127
AfricanAmerican4	0.098546	-0.054732
Latino4	0.833767	0.265917
Islander4	0.176991	0.940405
White4	0.136859	0.378943
Multiracial4	0.045885	0.146847
LimitedEnglish4	0.293614	0.225757
Disadv4	0.417239	0.761421
AllMath4Tested	0.493930	0.575876
AllMath4	0.463463	0.793577
NativeMath4	0.025953	0.080432
AfricanAmericanMath4	0.161130	-0.077146
LatinoMath4	1.000000	0.166465
IslanderMath4	0.166465	1.000000
WhiteMath4	0.132373	0.382159
MultiracialMath4	0.066978	0.145016
LimitedEnglishMath4	0.180779	0.610268
DisadvMath4	0.494340	0.788233
Count of Students in HS Admissions	0.352206	0.640299
Count of Testers	0.331389	0.790181
Count of Offers	0.157090	0.781842

	WhiteMath4	MultiracialMath4 \
SED Code	0.111477	-0.100134
Economic Need Index	-0.566390	-0.266687
Average ELA Proficiency	0.568440	0.327528
Average Math Proficiency	0.501042	0.284416
Number of students who registered for the SHSAT	0.101141	0.001986
Number of students who took the SHSAT	0.291501	0.167718
AllTested	0.402895	0.133575
All4	0.779631	0.404398

Native4	-0.011612	-0.010621
AfricanAmerican4	-0.004120	0.110634
Latino4	0.249102	0.121022
Islander4	0.443344	0.189189
White4	0.973047	0.488634
Multiracial4	0.447110	0.903375
LimitedEnglish4	0.009921	-0.034339
Disadv4	0.473337	0.135630
AllMath4Tested	0.386273	0.113615
AllMath4	0.700406	0.348437
NativeMath4	-0.010641	-0.012548
AfricanAmericanMath4	-0.048821	0.094581
LatinoMath4	0.132373	0.066978
IslanderMath4	0.382159	0.145016
WhiteMath4	1.000000	0.466884
MultiracialMath4	0.466884	1.000000
LimitedEnglishMath4	0.065524	-0.015748
DisadvMath4	0.388335	0.109534
Count of Students in HS Admissions	0.443029	0.130456
Count of Testers	0.643815	0.302159
Count of Offers	0.727092	0.532177

	LimitedEnglishMath4	\
SED Code	0.101698	
Economic Need Index	-0.015378	
Average ELA Proficiency	0.124413	
Average Math Proficiency	0.224012	
Number of students who registered for the SHSAT	0.221893	
Number of students who took the SHSAT	0.013616	
AllTested	0.358098	
All4	0.254364	
Native4	0.002829	
AfricanAmerican4	-0.059808	
Latino4	0.166169	
Islander4	0.399347	
White4	0.053820	
Multiracial4	-0.019669	
LimitedEnglish4	0.379997	
Disadv4	0.328336	
AllMath4Tested	0.383147	
AllMath4	0.427443	
NativeMath4	0.007294	
AfricanAmericanMath4	-0.042476	
LatinoMath4	0.180779	
IslanderMath4	0.610268	
WhiteMath4	0.065524	
MultiracialMath4	-0.015748	
LimitedEnglishMath4	1.000000	

DisadvMath4	0.491741
Count of Students in HS Admissions	0.458174
Count of Testers	0.441553
Count of Offers	0.286620

	DisadvMath4 \
SED Code	0.172113
Economic Need Index	-0.209139
Average ELA Proficiency	0.427474
Average Math Proficiency	0.517739
Number of students who registered for the SHSAT	0.102049
Number of students who took the SHSAT	0.071933
AllTested	0.697246
All4	0.746844
Native4	0.086700
AfricanAmerican4	0.294488
Latino4	0.506948
Islander4	0.746271
White4	0.377947
Multiracial4	0.067370
LimitedEnglish4	0.233133
Disadv4	0.906080
AllMath4Tested	0.707533
AllMath4	0.879551
NativeMath4	0.110006
AfricanAmericanMath4	0.341562
LatinoMath4	0.494340
IslanderMath4	0.788233
WhiteMath4	0.388335
MultiracialMath4	0.109534
LimitedEnglishMath4	0.491741
DisadvMath4	1.000000
Count of Students in HS Admissions	0.611494
Count of Testers	0.732673
Count of Offers	0.599504

	Count of Students in HS Admissions
SED Code	0.355074
Economic Need Index	-0.252197
Average ELA Proficiency	0.210394
Average Math Proficiency	0.194387
Number of students who registered for the SHSAT	0.058574
Number of students who took the SHSAT	0.144917
AllTested	0.909962
All4	0.651056
Native4	0.204923
AfricanAmerican4	0.055765
Latino4	0.578781

Islander4	0.616307
White4	0.449475
Multiracial4	0.103813
LimitedEnglish4	0.230929
Disadv4	0.692586
AllMath4Tested	0.897209
AllMath4	0.612487
NativeMath4	0.198148
AfricanAmericanMath4	-0.054994
LatinoMath4	0.352206
IslanderMath4	0.640299
WhiteMath4	0.443029
MultiracialMath4	0.130456
LimitedEnglishMath4	0.458174
DisadvMath4	0.611494
Count of Students in HS Admissions	1.000000
Count of Testers	0.856769
Count of Offers	0.521510

	Count of Testers \
SED Code	0.281210
Economic Need Index	-0.411867
Average ELA Proficiency	0.465485
Average Math Proficiency	0.442862
Number of students who registered for the SHSAT	0.109380
Number of students who took the SHSAT	0.333936
AllTested	0.804685
All4	0.852334
Native4	0.107553
AfricanAmerican4	0.134114
Latino4	0.541865
Islander4	0.780939
White4	0.645959
Multiracial4	0.281024
LimitedEnglish4	0.194009
Disadv4	0.807129
AllMath4Tested	0.782446
AllMath4	0.813443
NativeMath4	0.122513
AfricanAmericanMath4	0.018657
LatinoMath4	0.331389
IslanderMath4	0.790181
WhiteMath4	0.643815
MultiracialMath4	0.302159
LimitedEnglishMath4	0.441553
DisadvMath4	0.732673
Count of Students in HS Admissions	0.856769
Count of Testers	1.000000

Count of Offers	0.778770
-----------------	----------

	Count of Offers
SED Code	0.119194
Economic Need Index	-0.421596
Average ELA Proficiency	0.520695
Average Math Proficiency	0.468433
Number of students who registered for the SHSAT	0.101141
Number of students who took the SHSAT	0.291501
AllTested	0.485583
All4	0.859680
Native4	0.038949
AfricanAmerican4	0.000243
Latino4	0.285010
Islander4	0.829191
White4	0.738594
Multiracial4	0.542744
LimitedEnglish4	0.103076
Disadv4	0.678692
AllMath4Tested	0.456146
AllMath4	0.792829
NativeMath4	0.061064
AfricanAmericanMath4	-0.044907
LatinoMath4	0.157090
IslanderMath4	0.781842
WhiteMath4	0.727092
MultiracialMath4	0.532177
LimitedEnglishMath4	0.286620
DisadvMath4	0.599504
Count of Students in HS Admissions	0.521510
Count of Testers	0.778770
Count of Offers	1.000000

[29 rows x 29 columns]

1.2.2 Exploratory Visualization

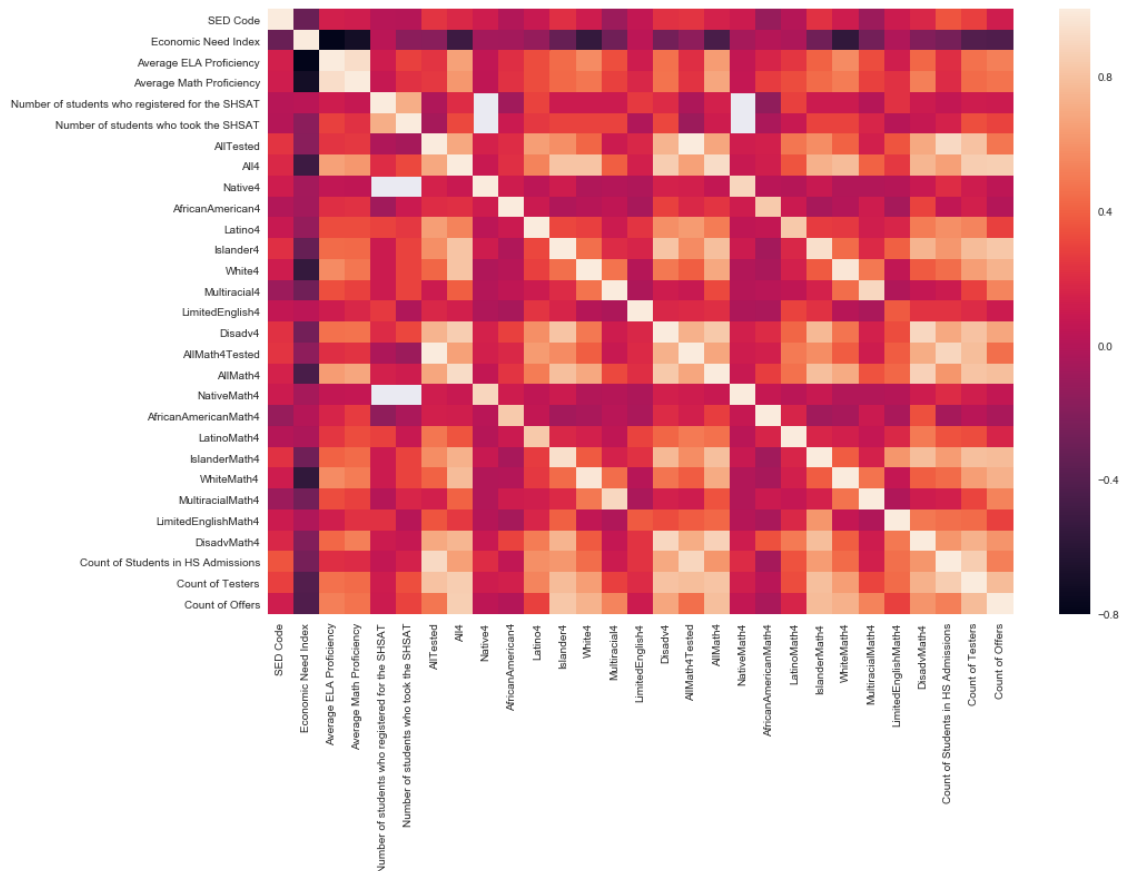
```
In [598]: # Now we use corr() to get the feature correlations and
          #then visualize them using a heatmap
          #The correlation values are fed into the heatmap to gain further insight.

          # Documentation referred at:
          # http://seaborn.pydata.org/generated/seaborn.heatmap.html
          import numpy as np; np.random.seed(0)
          import seaborn as sns; sns.set()
          import matplotlib.pyplot as plt

          plt.subplots(figsize=(15,10))
```

```
sns.heatmap(df_sub.corr())
```

```
Out[598]: <matplotlib.axes._subplots.AxesSubplot at 0x1a2ad05828>
```



1.2.3 Insights:

From the correlation values and heatmap, we can see that 'Economic Need Index' is inversely correlated with high negative values to - 'Average ELA Proficiency', - 'Average Math Proficiency', - 'White4', - 'WhiteMath4', - 'Count of Testers' and - 'Count of Offers'. Thus, the general assumption that schools with greater economic needs - perform more poorly on ELA and Math tests, - have lesser White students, - have fewer students taking the tests and - consequently getting less SHSAT offers holds true.

Now to apply clustering techniques, we will limit the number of features further. We will use only the following columns as : - 'Economic Need Index', - 'AfricanAmerican4', - 'Latino4', - 'White4', - 'AfricanAmericanMath4', - 'LatinoMath4', - 'WhiteMath4', - 'Number of students who took the SHSAT', - 'Count of Students in HS Admissions', - 'Count of Offers']

```
In [599]: sub = df_sub[['Economic Need Index',  
                        'AfricanAmerican4',
```

```

    'Latino4',
    'White4',
    'AfricanAmericanMath4',
    'LatinoMath4',
    'WhiteMath4',
    'Number of students who took the SHSAT',
    'Count of Offers']]
sub.info()

```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Int64Index: 1364 entries, 0 to 1363
```

```
Data columns (total 9 columns):
```

Economic Need Index	1339 non-null float64
AfricanAmerican4	1364 non-null float64
Latino4	1364 non-null float64
White4	1364 non-null float64
AfricanAmericanMath4	1364 non-null float64
LatinoMath4	1364 non-null float64
WhiteMath4	1364 non-null float64
Number of students who took the SHSAT	113 non-null float64
Count of Offers	683 non-null float64

```

dtypes: float64(9)
memory usage: 106.6 KB

```

```
In [600]: sub.corr()
```

```
Out[600]:
```

	Economic Need Index	AfricanAmerican4	\
Economic Need Index	1.000000	-0.072707	
AfricanAmerican4	-0.072707	1.000000	
Latino4	-0.127446	0.095225	
White4	-0.549216	0.012959	
AfricanAmericanMath4	0.004962	0.839716	
LatinoMath4	-0.029510	0.098546	
WhiteMath4	-0.566390	-0.004120	
Number of students who took the SHSAT	-0.172916	0.098985	
Count of Offers	-0.421596	0.000243	

	Latino4	White4	\
Economic Need Index	-0.127446	-0.549216	
AfricanAmerican4	0.095225	0.012959	
Latino4	1.000000	0.277335	
White4	0.277335	1.000000	
AfricanAmericanMath4	0.058003	-0.044113	
LatinoMath4	0.833767	0.136859	
WhiteMath4	0.249102	0.973047	
Number of students who took the SHSAT	0.254371	0.291501	
Count of Offers	0.285010	0.738594	

	AfricanAmericanMath4	LatinoMath4 \
Economic Need Index	0.004962	-0.029510
AfricanAmerican4	0.839716	0.098546
Latino4	0.058003	0.833767
White4	-0.044113	0.136859
AfricanAmericanMath4	1.000000	0.161130
LatinoMath4	0.161130	1.000000
WhiteMath4	-0.048821	0.132373
Number of students who took the SHSAT	-0.036843	0.080072
Count of Offers	-0.044907	0.157090

	WhiteMath4 \
Economic Need Index	-0.566390
AfricanAmerican4	-0.004120
Latino4	0.249102
White4	0.973047
AfricanAmericanMath4	-0.048821
LatinoMath4	0.132373
WhiteMath4	1.000000
Number of students who took the SHSAT	0.291501
Count of Offers	0.727092

	Number of students who took the SHSAT \
Economic Need Index	-0.172916
AfricanAmerican4	0.098985
Latino4	0.254371
White4	0.291501
AfricanAmericanMath4	-0.036843
LatinoMath4	0.080072
WhiteMath4	0.291501
Number of students who took the SHSAT	1.000000
Count of Offers	0.291501

	Count of Offers
Economic Need Index	-0.421596
AfricanAmerican4	0.000243
Latino4	0.285010
White4	0.738594
AfricanAmericanMath4	-0.044907
LatinoMath4	0.157090
WhiteMath4	0.727092
Number of students who took the SHSAT	0.291501
Count of Offers	1.000000

```
In [601]: sub=sub.fillna(0)
```

```
In [602]: # copy dataframe into sub1 for decision tree to get score
```

```

sub1 = sub
sub1.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 1364 entries, 0 to 1363
Data columns (total 9 columns):
Economic Need Index          1364 non-null float64
AfricanAmerican4            1364 non-null float64
Latino4                      1364 non-null float64
White4                      1364 non-null float64
AfricanAmericanMath4        1364 non-null float64
LatinoMath4                  1364 non-null float64
WhiteMath4                   1364 non-null float64
Number of students who took the SHSAT 1364 non-null float64
Count of Offers              1364 non-null float64
dtypes: float64(9)
memory usage: 106.6 KB

```

We use Decision Tree method to determine if indeed our selected features are relevant or not using score() methods.

```

In [603]: from sklearn.tree import DecisionTreeRegressor
          from sklearn.cross_validation import train_test_split

          # Done: Make a copy of the DataFrame, using the 'drop' function to drop the given feature
          new_data = sub1.drop('Count of Offers', axis=1)

          # Done: Split the data into training and testing sets(0.25) using the given feature
          # Set a random state.
          X_train, X_test, y_train, y_test = train_test_split(new_data, sub1['Count of Offers'])

          # Done: Create a decision tree regressor and fit it to the training set
          regressor = DecisionTreeRegressor(random_state=32)
          regressor.fit(X_train,y_train)

          # Done: Report the score of the prediction using the testing set
          score = regressor.score(X_test,y_test)

          print(round(score,4))

```

0.0923

```

In [604]: # copy dataframe into sub2 for decision tree to get score
          sub2 = sub
          sub2.info()

          from sklearn.tree import DecisionTreeRegressor

```

```

from sklearn.cross_validation import train_test_split

# Done: Make a copy of the DataFrame, using the 'drop' function to drop the given feature
new_data = sub2.drop('Economic Need Index', axis=1)

# Done: Split the data into training and testing sets(0.25) using the given feature
# Set a random state.
X_train, X_test, y_train, y_test = train_test_split(new_data, sub2['Economic Need Index'],
                                                    test_size=0.25, random_state=32)

# Done: Create a decision tree regressor and fit it to the training set
regressor = DecisionTreeRegressor(random_state=32)
regressor.fit(X_train,y_train)

# Done: Report the score of the prediction using the testing set
score = regressor.score(X_test,y_test)

print(round(score,4))

```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 1364 entries, 0 to 1363
Data columns (total 9 columns):
Economic Need Index      1364 non-null float64
AfricanAmerican4         1364 non-null float64
Latino4                  1364 non-null float64
White4                   1364 non-null float64
AfricanAmericanMath4     1364 non-null float64
LatinoMath4              1364 non-null float64
WhiteMath4               1364 non-null float64
Number of students who took the SHSAT  1364 non-null float64
Count of Offers          1364 non-null float64
dtypes: float64(9)
memory usage: 146.6 KB
0.2061

```

```

In [605]: # copy dataframe into sub3 for decision tree to get score
sub3 = sub
sub3.info()

```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 1364 entries, 0 to 1363
Data columns (total 9 columns):
Economic Need Index      1364 non-null float64
AfricanAmerican4         1364 non-null float64
Latino4                  1364 non-null float64
White4                   1364 non-null float64
AfricanAmericanMath4     1364 non-null float64
LatinoMath4              1364 non-null float64

```

```

WhiteMath4                1364 non-null float64
Number of students who took the SHSAT    1364 non-null float64
Count of Offers            1364 non-null float64
dtypes: float64(9)
memory usage: 146.6 KB

```

```

In [606]: from sklearn.tree import DecisionTreeRegressor
          from sklearn.cross_validation import train_test_split

          # Done: Make a copy of the DataFrame, using the 'drop' function to drop the given feature
          new_data = sub3.drop('Number of students who took the SHSAT', axis=1)

          # Done: Split the data into training and testing sets(0.25) using the given feature
          # Set a random state.
          X_train, X_test, y_train, y_test = train_test_split(new_data, sub3['Number of students who took the SHSAT'],
                                                              test_size=0.25, random_state=32)

          # Done: Create a decision tree regressor and fit it to the training set
          regressor = DecisionTreeRegressor(random_state=32)
          regressor.fit(X_train,y_train)

          # Done: Report the score of the prediction using the testing set
          score = regressor.score(X_test,y_test)

          print(round(score,4))

0.3862

```

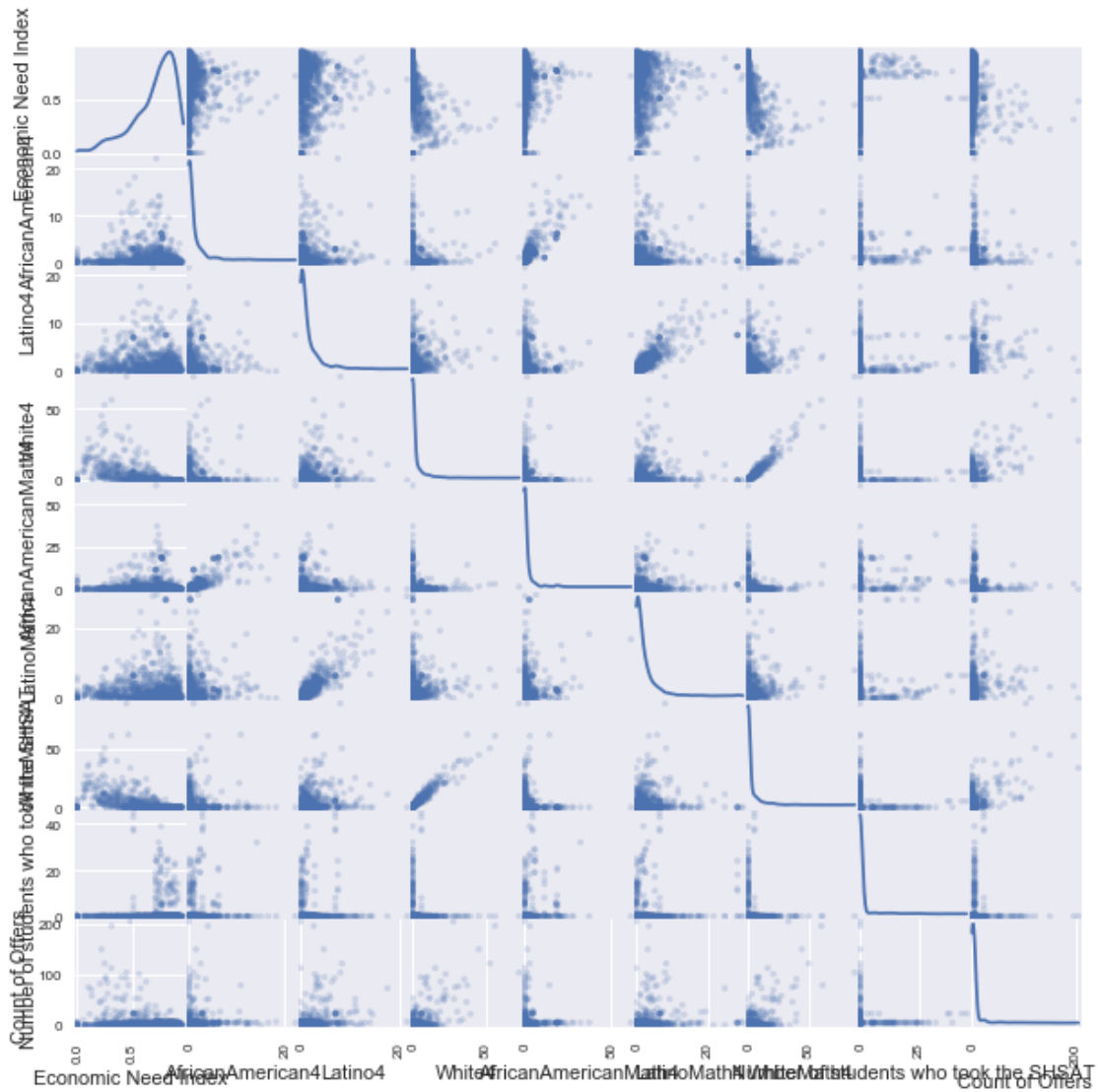
The coefficient of determination, R^2 , is scored between 0 and 1, with 1 being a perfect fit. A negative R^2 implies the model fails to fit the data. If you get a low score for a particular feature, that lends us to believe that that feature point is hard to predict using the other features, thereby making it an important feature to consider when considering relevance. Thus, 'Count of Offers' and 'Economic Need Index' are important features while 'Number of students who took the SHSAT' is of medium importance and validate our choice of features in the subset we are considering.

1.2.4 Algorithms and Techniques

Here we will be using Kmeans and GMM algorithms to determine the best clustering method for the datasets. We will use scatterplot matrix, Elbow curve and Silhouette score to evaluate if the number of clusters is optimal. The silhouette score will also be used to compare against benchmark model. We are using Kmeans clustering as it is easy to implement, provides tight clusters and with our use of Silhouette score can give us the optimal number of clusters as well. (offsets its disadvantage of difficulty in predicting number of clusters) The Gaussian mixture model (GMM) attempts to find a mixture of multi-dimensional Gaussian probability distributions that best model any input dataset. This method will also ensure that the main practical issues with k-means will be removed namely, allowing for a full covariance and not using hard cutoffs for cluster assignment within the training set. thus, we can be sure that the best clusters and numbers are obtained.


```
In [607]: from pandas.plotting import scatter_matrix

scatter_matrix(sub, alpha=0.2, figsize=(10, 10), diagonal='kde');
```



The Elbow method is a method of interpretation and validation of consistency within cluster analysis designed to help finding the appropriate number of clusters in a dataset.

```
In [608]: import pylab as pl
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import MinMaxScaler

#Scaling of data
ss = StandardScaler()
ss.fit_transform(sub)
```

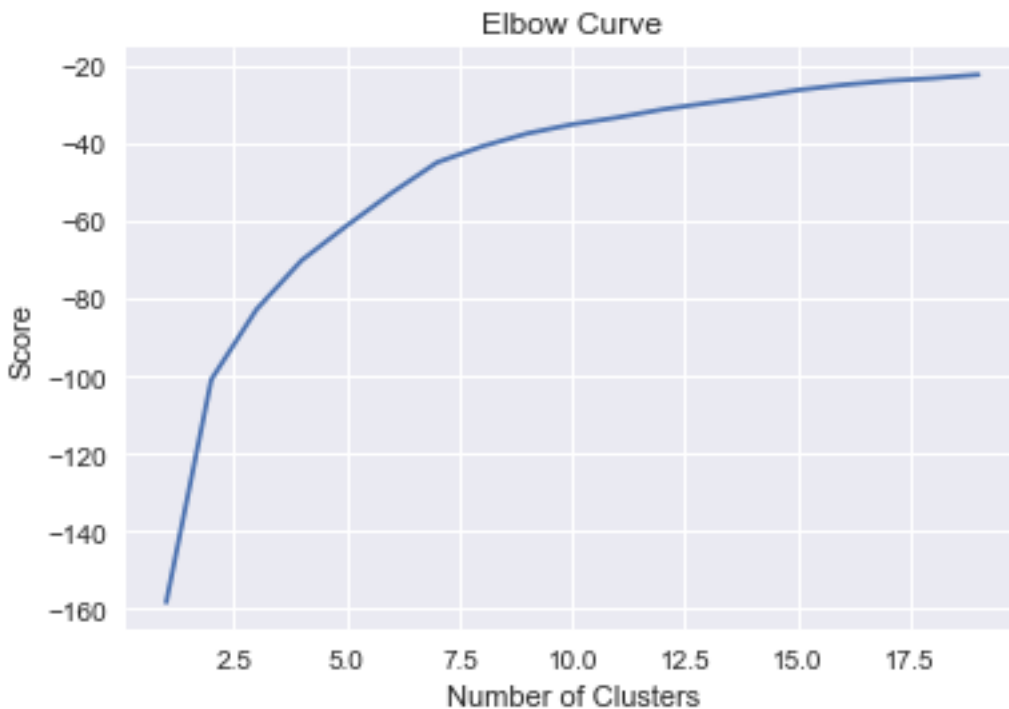
```

minmax = MinMaxScaler()
minmax.fit_transform(sub)

x = sub.iloc[:, ].values
z = minmax.fit_transform(x)
Nc = range(1, 20)
kmeans = [KMeans(n_clusters=i) for i in Nc]

score = [kmeans[i].fit(z).score(z) for i in range(len(kmeans))]
pl.plot(Nc,score)
pl.xlabel('Number of Clusters')
pl.ylabel('Score')
pl.title('Elbow Curve')
pl.show()

```



Thus, we can consider 3 clusters although 7 or 8 clusters also look feasible using the Elbow Curve method.

1.3 Methodology

1.3.1 Data Preprocessing

Since the dataset is from Kaggle, only minimal preprocessing was done on the dataset. Most of the wrangling was simply joining the datasets and subsetting to include only relevant attributes. The

rating percentages had converted to float values with the percent symbol stripped while test scores for all each particular race was computed as average of individual grade level scores. Feature relevance was determined by using creating a decision tree regressor and fitting it to the training set and then reporting the score of the prediction using the testing set.

1.3.2 Implementation

For execution, we will be using sklearn to determine the clusters and centers using K-means. We will then do the same with GMM algorithm and then compute the silhouette score. We shall do so for cluster numbered for 3,4,7,8. We will then compute silhouette score for clusters numbered 3 to 8 using Kmeans and additionally create cluster maps to see how each cluster number chosen fared.

1.3.3 Refinement

The refinement was basically to evaluate which cluster number gave the best results. Hence, both GMM and Kmeans algorithms were used. Also, my proposal simply considered silhouette score but I added using elbow curve method as well and evaluated this metric for all the cluster numbers and for both algorithms to hone in on the right groups.

Section 1.5.3: Back to Top

1.3.4 Cluster centers for n=8 clusters using K-means

```
In [609]: from sklearn.cluster import KMeans
import numpy as np
X = sub
kmeans = KMeans(n_clusters=8, random_state=0).fit(sub)
kmeans.labels_
#kmeans.predict([[0, 0], [4, 4]])
cluster_centers = kmeans.cluster_centers_
print(cluster_centers)
```

```
[[ 2.86024096e-01  3.19277108e-01  1.63253012e+00  1.14959839e+01
   3.19277108e-01  2.05823293e+00  1.50582329e+01  0.00000000e+00
   2.16867470e+00]
 [ 3.41000000e-01  1.80000000e+00  4.73333333e+00  5.07000000e+01
   1.86666667e+00  4.60000000e+00  5.52333333e+01  0.00000000e+00
   1.55400000e+02]
 [ 7.03391595e-01  6.28621458e-01  8.40337472e-01  4.89652977e-01
   7.32569245e-01  1.19134034e+00  6.72715696e-01  1.43266476e-01
   2.26552053e+00]
 [ 3.41120000e-01  1.22000000e+00  4.58000000e+00  1.92600000e+01
   1.16666667e+00  4.87333333e+00  2.15400000e+01 -2.22044605e-16
   7.18800000e+01]
 [ 5.21208333e-01  1.37847222e+00  4.88888889e+00  5.78125000e+00
   1.14236111e+00  5.01041667e+00  5.55208333e+00  3.33333333e-01
   2.34166667e+01]
```

```
[ 7.61400000e-01  1.67000000e+00  1.29000000e+00  4.80000000e-01
 3.86333333e+00  2.12333333e+00  4.13333333e-01  2.05400000e+01
 6.44000000e+00]
[ 6.85725490e-01  7.95751634e+00  1.26470588e+00  7.84313725e-02
 1.78790850e+01  3.06209150e+00  8.16993464e-02  1.15686275e+00
 4.62745098e+00]
[ 7.10927273e-01  9.93939394e-01  6.83030303e+00  9.54545455e-01
 1.63939394e+00  1.27515152e+01  1.11212121e+00  7.45454545e-01
 2.65454545e+00]]
```

1.3.5 Silhouette score for n=8 clusters using GMM

```
In [610]: # Links used:
# http://scikit-learn.org/stable/modules/clustering.html
# http://scikit-learn.org/stable/modules/generated/sklearn.metrics.silhouette_score.
from sklearn.mixture import GMM
from sklearn.metrics import silhouette_score
import warnings
warnings.filterwarnings('ignore')

# Done: Apply your clustering algorithm of choice to the reduced data
x3 = GMM(n_components=8)

reduced_data=sub
clusterer = x3.fit(reduced_data)

# Done: Predict the cluster for each data point
preds = clusterer.predict(reduced_data)

# Done: Find the cluster centers
centers = clusterer.means_

# Done: Predict the cluster for each transformed sample data point
sample_preds = clusterer.predict(cluster_centers)

# Done: Calculate the mean silhouette coefficient for the number of clusters chosen
score = silhouette_score(reduced_data, preds)

print(round(score,4))
print("Scores for 8 clusters is {}".format(round(score,4)))

0.0867
Scores for 8 clusters is 0.0867
```

1.3.6 Cluster centers for n=3 clusters using K-means

```
In [611]: from sklearn.cluster import KMeans
import numpy as np
X2 = sub
kmeans = KMeans(n_clusters=3, random_state=0).fit(X2)
kmeans.labels_
cluster_centers = kmeans.cluster_centers_
print(cluster_centers)

[[ 6.73086336e-01  9.71596597e-01  1.31481481e+00  1.34034034e+00
  1.53365866e+00  1.96559059e+00  1.70170170e+00  9.70720721e-01
  3.23198198e+00]
 [ 3.45518519e-01  1.15432099e+00  4.36419753e+00  1.92962963e+01
  1.10493827e+00  4.66666667e+00  2.14259259e+01 -3.33066907e-16
  6.93333333e+01]
 [ 3.41000000e-01  1.80000000e+00  4.73333333e+00  5.07000000e+01
  1.86666667e+00  4.60000000e+00  5.52333333e+01  0.00000000e+00
  1.55400000e+02]]
```

1.3.7 Silhouette score for n=3 clusters using GMM

```
In [612]: # Links used:
# http://scikit-learn.org/stable/modules/clustering.html
# http://scikit-learn.org/stable/modules/generated/sklearn.metrics.silhouette_score.
from sklearn.mixture import GMM
from sklearn.metrics import silhouette_score
import warnings
warnings.filterwarnings('ignore')

# Done: Apply your clustering algorithm of choice to the reduced data
x3 = GMM(n_components=3)

reduced_data2=sub
clusterer = x3.fit(reduced_data2)

# Done: Predict the cluster for each data point
preds = clusterer.predict(reduced_data2)

# Done: Find the cluster centers
centers = clusterer.means_

# Done: Predict the cluster for each transformed sample data point
sample_preds = clusterer.predict(centers)

# Done: Calculate the mean silhouette coefficient for the number of clusters chosen
score = silhouette_score(reduced_data2, preds)
```

```
print(round(score,4))
print("Scores for 3 clusters is {}".format(round(score,4)))
```

0.2614

Scores for 3 clusters is 0.2614

1.3.8 Cluster centers for n=7 clusters using K-means

```
In [613]: from sklearn.cluster import KMeans
import numpy as np
X2 = sub
kmeans = KMeans(n_clusters=7, random_state=0).fit(X2)
kmeans.labels_
cluster_centers = kmeans.cluster_centers_
print(cluster_centers)
```

```
[[ 6.88205457e-01  4.97057250e-01  9.84216158e-01  6.23863028e-01
  7.41305511e-01  1.76404494e+00  9.21883360e-01 -4.44089210e-16
  5.50670620e-14]
 [ 3.41000000e-01  1.80000000e+00  4.73333333e+00  5.07000000e+01
  1.86666667e+00  4.60000000e+00  5.52333333e+01  0.00000000e+00
  1.55400000e+02]
 [ 3.41120000e-01  1.22000000e+00  4.58000000e+00  1.92600000e+01
  1.16666667e+00  4.87333333e+00  2.15400000e+01 -2.22044605e-16
  7.18800000e+01]
 [ 5.19595745e-01  1.40780142e+00  4.94680851e+00  5.85460993e+00
  1.16666667e+00  5.09574468e+00  5.62765957e+00  3.40425532e-01
  2.36170213e+01]
 [ 7.20851562e-01  8.88671875e-01  1.24479167e+00  4.40429688e-01
  9.38802083e-01  1.51660156e+00  5.08463542e-01  7.81250000e-01
  5.19921875e+00]
 [ 2.77413333e-01  3.44444444e-01  1.63777778e+00  1.21000000e+01
  3.40000000e-01  2.02000000e+00  1.57688889e+01  4.44089210e-16
  2.13333333e+00]
 [ 7.05155844e-01  5.69047619e+00  1.93290043e+00  3.63636364e-01
  1.32554113e+01  4.62121212e+00  3.22510823e-01  1.13896104e+01
  5.81818182e+00]]
```

1.3.9 Silhouette score for n=7 clusters using GMM

```
In [614]: # Links used:
# http://scikit-learn.org/stable/modules/clustering.html
# http://scikit-learn.org/stable/modules/generated/sklearn.metrics.silhouette_score.
from sklearn.mixture import GMM
from sklearn.metrics import silhouette_score
import warnings
```

```
warnings.filterwarnings('ignore')

# Done: Apply your clustering algorithm of choice to the reduced data
x3 = GMM(n_components=7)

reduced_data2=sub
clusterer = x3.fit(reduced_data2)

# Done: Predict the cluster for each data point
preds = clusterer.predict(reduced_data2)

# Done: Find the cluster centers
centers = clusterer.means_

# Done: Predict the cluster for each transformed sample data point
sample_preds = clusterer.predict(cluster_centers)

# Done: Calculate the mean silhouette coefficient for the number of clusters chosen
score = silhouette_score(reduced_data2, preds)

print(round(score,4))
print("Scores for 7 clusters is {}".format(round(score,4)))
```

0.1008

Scores for 7 clusters is 0.1008

1.3.10 Cluster centers for n=4 clusters using K-means

```
In [615]: from sklearn.cluster import KMeans
import numpy as np
X2 = sub
kmeans = KMeans(n_clusters=4, random_state=0).fit(X2)
kmeans.labels_
cluster_centers = kmeans.cluster_centers_
print(cluster_centers)
```

```
[[ 7.01517270e-01  9.85882675e-01  1.12595943e+00  5.18777412e-01
  1.60389254e+00  1.80317982e+00  7.12856360e-01  9.16118421e-01
  2.55016447e+00]
 [ 3.49615385e-01  1.19230769e+00  4.49358974e+00  1.85256410e+01
  1.14743590e+00  4.81410256e+00  2.07179487e+01 -3.33066907e-16
  7.11538462e+01]
 [ 3.41000000e-01  1.80000000e+00  4.73333333e+00  5.07000000e+01
  1.86666667e+00  4.60000000e+00  5.52333333e+01  0.00000000e+00
  1.55400000e+02]
 [ 3.73888889e-01  8.16239316e-01  3.27492877e+00  1.02037037e+01
```

```
7.90598291e-01  3.64387464e+00  1.23048433e+01  1.52991453e+00
1.04786325e+01]]
```

1.3.11 Silhouette score for n=4 clusters using GMM

```
In [616]: # Links used:
# http://scikit-learn.org/stable/modules/clustering.html
# http://scikit-learn.org/stable/modules/generated/sklearn.metrics.silhouette_score.
from sklearn.mixture import GMM
from sklearn.metrics import silhouette_score
import warnings
warnings.filterwarnings('ignore')

# Done: Apply your clustering algorithm of choice to the reduced data
x3 = GMM(n_components=4)

reduced_data2=sub
clusterer = x3.fit(reduced_data2)

# Done: Predict the cluster for each data point
preds = clusterer.predict(reduced_data2)

# Done: Find the cluster centers
centers = clusterer.means_

# Done: Predict the cluster for each transformed sample data point
sample_preds = clusterer.predict(cluster_centers)

# Done: Calculate the mean silhouette coefficient for the number of clusters chosen
score = silhouette_score(reduced_data2, preds)

print(round(score,4))
print("Scores for 4 clusters is {}".format(round(score,4)))

0.1711
Scores for 4 clusters is 0.1711
```

```
In [617]: #Creating cluster map to visualize clusters
from sklearn.metrics import silhouette_samples, silhouette_score
```

```
In [618]: X = sub[['Economic Need Index',
'AfricanAmerican4',
'Latino4',
'White4',
'AfricanAmericanMath4',
'LatinoMath4',
```



```

'WhiteMath4',
'Number of students who took the SHSAT',
'Count of Offers']]

```

```

from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
X_scaled = scaler.fit_transform( X )

```

```

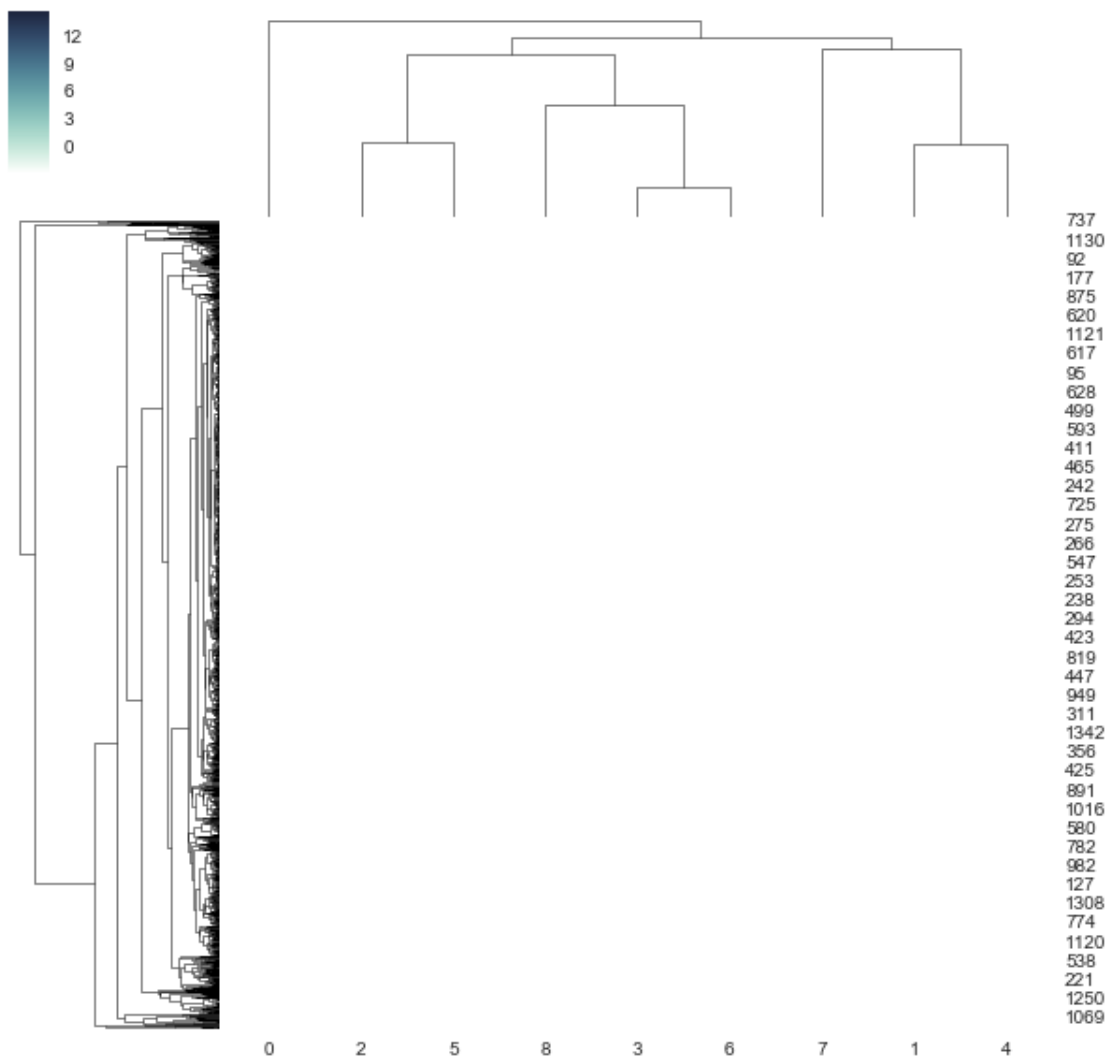
In [619]: import matplotlib.cm as cm
import seaborn as sn
cmap = sn.cubehelix_palette(as_cmap=True, rot=-.3, light=1)

```

```

In [620]: sn.clustermap(X_scaled, cmap=cmap, linewidths=.5);

```



1.3.12 Cluster maps and silhouette scores using Kmeans for number of clusters ranging from 3 to 8

In [621]: %matplotlib inline

```
cluster_range = range( 3, 9 ) #range excludes upper bound

for n_clusters in cluster_range:
    fig, (ax1, ax2) = plt.subplots(1, 2)
    fig.set_size_inches(18, 7)

    # The 1st subplot is the silhouette plot
    # The silhouette coefficient can range from -1, 1
    ax1.set_xlim([-1, 1])

    # The (n_clusters+1)*10 is for inserting blank space between silhouette
    # plots of individual clusters, to demarcate them clearly.
    ax1.set_ylim([0, len(X_scaled) + (n_clusters + 1) * 10])

    cluster_labels = clusterer.fit_predict( X_scaled )

    # The silhouette_score gives the average value for all the samples.
    # This gives a perspective into the density and separation of the formed
    # clusters
    silhouette_avg = silhouette_score(X_scaled, cluster_labels)
    print("For n_clusters =", n_clusters,
          "The average silhouette_score is :", silhouette_avg)

    # Compute the silhouette scores for each sample
    sample_silhouette_values = silhouette_samples(X_scaled, cluster_labels)

    y_lower = 10
    for i in range(n_clusters):
        # Aggregate the silhouette scores for samples belonging to
        # cluster i, and sort them
        ith_cluster_silhouette_values = \
            sample_silhouette_values[cluster_labels == i]

        ith_cluster_silhouette_values.sort()

        size_cluster_i = ith_cluster_silhouette_values.shape[0]
        y_upper = y_lower + size_cluster_i

        cmap = cm.get_cmap("Spectral")
        color = cmap(float(i) / n_clusters)

        #color = cm.spectral(float(i) / n_clusters)
        ax1.fill_betweenx(np.arange(y_lower, y_upper),
```

```

        0, ith_cluster_silhouette_values,
        facecolor=color, edgecolor=color, alpha=0.7)

    # Label the silhouette plots with their cluster numbers at the middle
    ax1.text(-0.05, y_lower + 0.5 * size_cluster_i, str(i))

    # Compute the new y_lower for next plot
    y_lower = y_upper + 10
    ax1.set_title("The silhouette plot for the various clusters.")
    ax1.set_xlabel("The silhouette coefficient values")
    ax1.set_ylabel("Cluster label")

    # The vertical line for average silhouette score of all the values
    ax1.axvline(x=silhouette_avg, color="red", linestyle="--")

    ax1.set_yticks([]) # Clear the yaxis labels / ticks
    ax1.set_xticks([-0.1, 0, 0.2, 0.4, 0.6, 0.8, 1])

    # 2nd Plot showing the actual clusters formed
    #cmap = cm.get_cmap("Spectral")
    colors = cmap(cluster_labels.astype(float) / n_clusters)
    #colors = cm.spectral(cluster_labels.astype(float) / n_clusters)
    ax2.scatter(X_scaled[:, 0], X_scaled[:, 1], marker='.', s=30, lw=0, alpha=0.7,
                c=colors)

    # Labeling the clusters
    centers = kmeans.cluster_centers_
    #centers = clusterer.cluster_centers_
    # Draw white circles at cluster centers
    ax2.scatter(centers[:, 0], centers[:, 1],
                marker='o', c="white", alpha=1, s=200)

    for i, c in enumerate(centers):
        ax2.scatter(c[0], c[1], marker='$%d$' % i, alpha=1, s=50)

    ax2.set_title("The visualization of the clustered data.")
    ax2.set_xlabel("Feature space for the 1st feature")
    ax2.set_ylabel("Feature space for the 2nd feature")

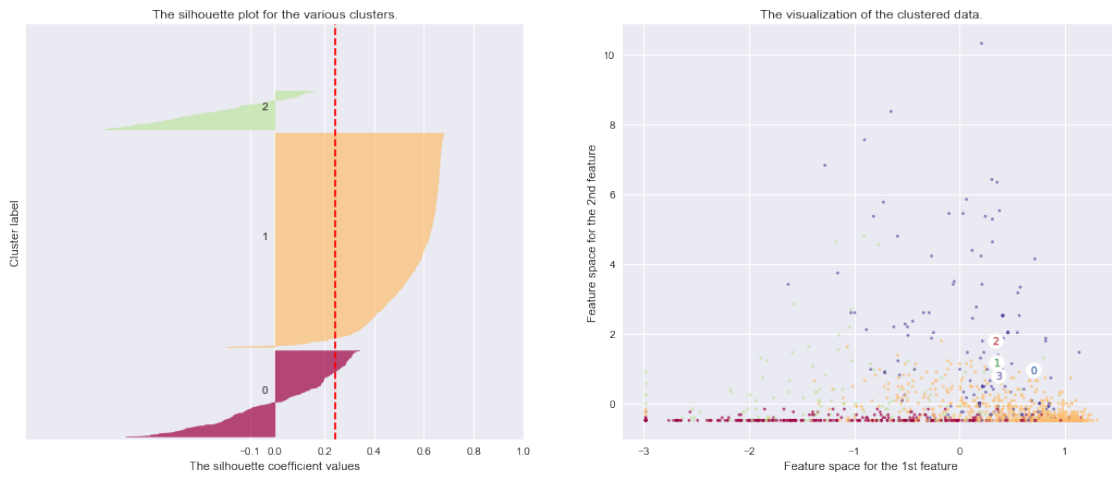
    plt.suptitle(("Silhouette analysis for KMeans clustering on sample data "
                  "with n_clusters = %d" % n_clusters),
                  fontsize=14, fontweight='bold')

    plt.show();

```

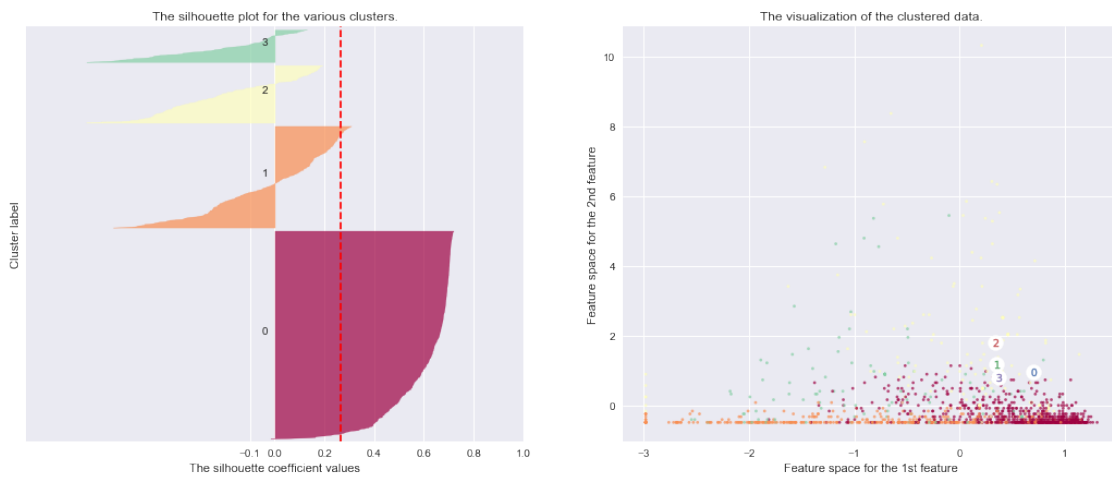
For n_clusters = 3 The average silhouette_score is : 0.243592391953

Silhouette analysis for KMeans clustering on sample data with n_clusters = 3

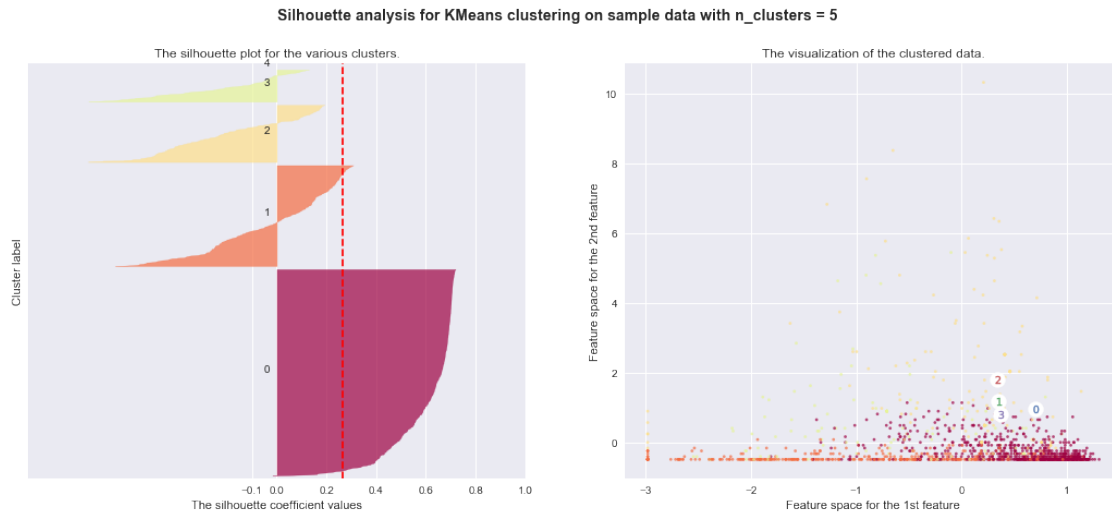


For n_clusters = 4 The average silhouette_score is : 0.266591843412

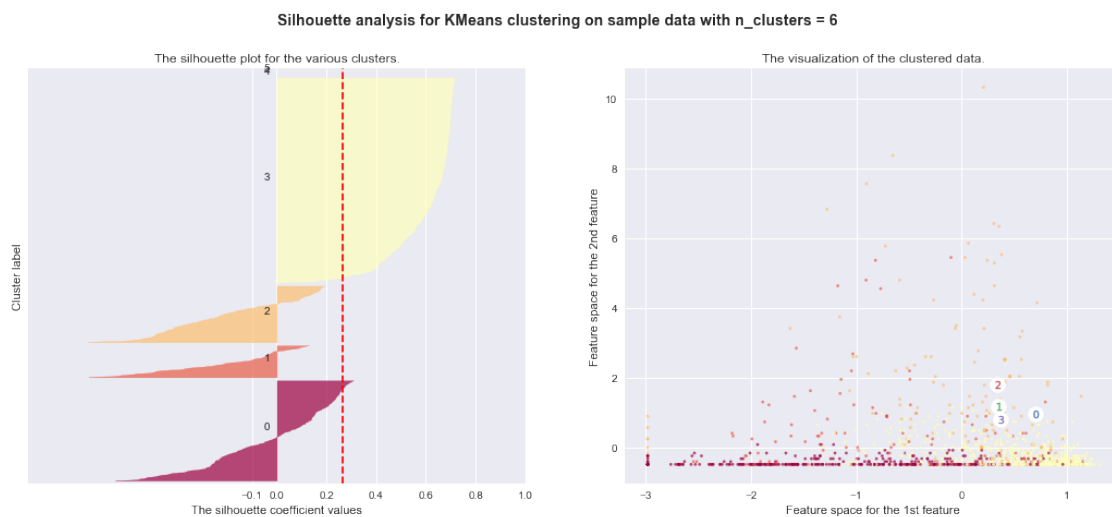
Silhouette analysis for KMeans clustering on sample data with n_clusters = 4



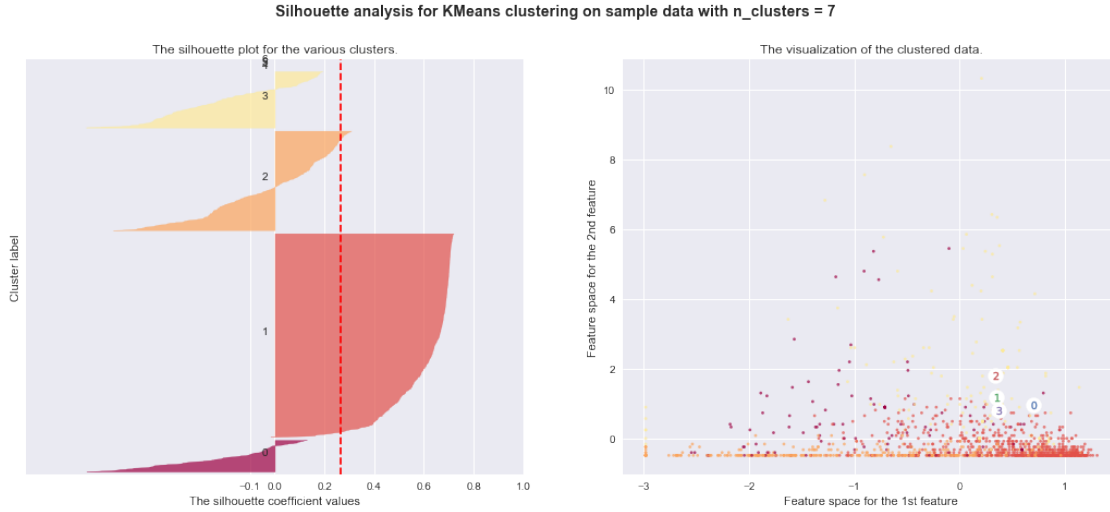
For n_clusters = 5 The average silhouette_score is : 0.266591843412



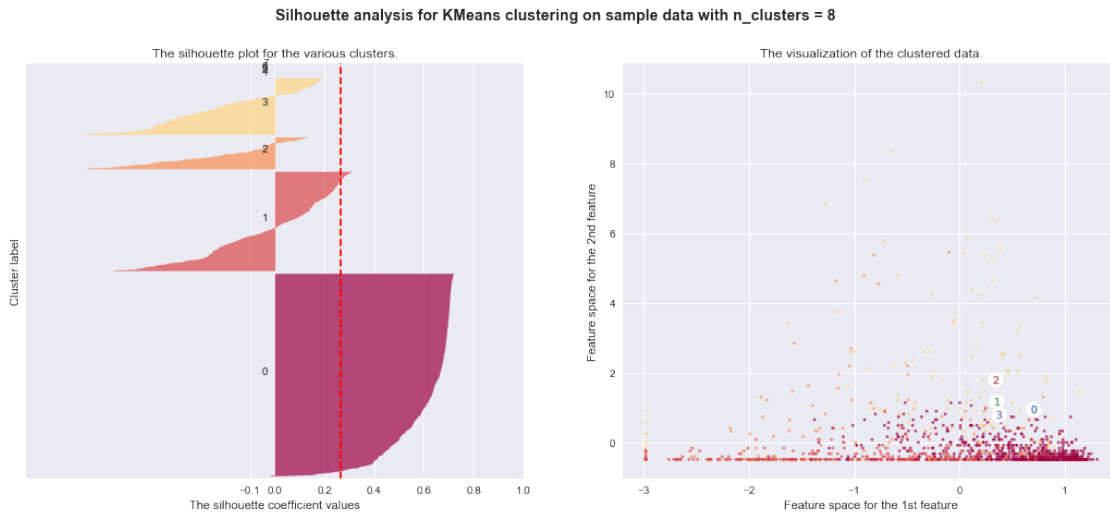
For n_clusters = 6 The average silhouette_score is : 0.266775421534



For n_clusters = 7 The average silhouette_score is : 0.266591843412



For `n_clusters = 8` The average `silhouette_score` is : 0.266591843412



Thus, for clusters =3, we get the best results when considering clustermap, GMM and K-means techniques.

1.3.13 Benchmark:

For this project, the benchmark model is the kernel from the Kaggle dataset at: <https://www.kaggle.com/laiyipeng/target-schools-action-recommended-for-passnyc> It was one of the winners when the competition (using this dataset) originally ran. There are 3 clusters present in this analysis. I shall try to compare my model with this analysis.

In the benchmark model, we have: 1) New York State annual test results 2) New York City Department of Education annual quality review 3) Other

Since this original kernel is in R and approaches the problem with a different angle and does not compute silhouette score, am reproducing the dataset and joins used and computing the silhouette scores for this model.

```
In [622]: benchmark =dfy
          benchmark.shape
```

```
Out[622]: (1364, 71)
```

```
In [623]: benchmark = benchmark.drop([
        'School Name',
        'SED Code',
        'Location Code',
        'District',
        'Latitude',
        'Longitude',
        'Address (Full)',
        'City',
        'Zip',
        'Grades',
        'Grade Low',
        'Grade High',
        'Community School?',
        'Economic Need Index',
        'School Income Estimate',
        'Percent ELL',
        'Percent Asian',
        'Percent Black',
        'Percent Hispanic',
        'Percent Black / Hispanic',
        'Percent White',
        'Student Achievement Rating',
        'Grade 3 Math 4s - Multiracial',
        'DBN',
        'School name',
        'Year of SHST',
        'Grade level',
        'Enrollment on 10/31',
        'Number of students who registered for the SHSAT',
        'Number of students who took the SHSAT',
        'AllTested',
        'All4',
        'Native4',
        'AfricanAmerican4',
        'Latino4',
        'Islander4',
```

```

        'White4',
        'Multiracial4',
        'LimitedEnglish4',
        'Disadv4',
        'AllMath4Tested',
        'AllMath4',
        'NativeMath4',
        'AfricanAmericanMath4',
        'LatinoMath4',
        'IslanderMath4',
        'WhiteMath4',
        'MultiracialMath4',
        'LimitedEnglishMath4',
        'DisadvMath4',
        'Feeder School DBN',
        'Feeder School Name',
        'Count of Students in HS Admissions',
        'Count of Testers',
        'Rigorous Instruction Rating',
        'Collaborative Teachers Rating',
        'Supportive Environment Rating',
        'Effective School Leadership Rating',
        'Strong Family-Community Ties Rating',
        'Trust Rating'],
axis=1)

```

```
In [624]: benchmark.shape
```

```
Out[624]: (1364, 11)
```

```
In [625]: list(benchmark)
```

```

Out[625]: ['Student Attendance Rate',
           'Percent of Students Chronically Absent',
           'Rigorous Instruction %',
           'Collaborative Teachers %',
           'Supportive Environment %',
           'Effective School Leadership %',
           'Strong Family-Community Ties %',
           'Trust %',
           'Average ELA Proficiency',
           'Average Math Proficiency',
           'Count of Offers']

```

```
In [626]: benchmark.head(2)
```

```

Out[626]:   Student Attendance Rate  Percent of Students Chronically Absent  \
0                94%                18%
1                92%                30%

```


	Rigorous Instruction %	Collaborative Teachers %	Supportive Environment %	\
0	89%	94%	86%	
1	96%	96%	97%	

	Effective School Leadership %	Strong Family-Community Ties %	Trust %	\
0	91%	85%	94%	
1	90%	86%	94%	

	Average ELA Proficiency	Average Math Proficiency	Count of Offers
0	2.14	2.17	NaN
1	2.63	2.98	NaN

```
In [627]: benchmark['Student Attendance Rate'] = benchmark['Student Attendance Rate'].str.rstrip()
benchmark['Percent of Students Chronically Absent'] = benchmark['Percent of Students Chronically Absent'].str.rstrip()
benchmark['Rigorous Instruction %'] = benchmark['Rigorous Instruction %'].str.rstrip()
benchmark['Collaborative Teachers %'] = benchmark['Collaborative Teachers %'].str.rstrip()
benchmark['Supportive Environment %'] = benchmark['Supportive Environment %'].str.rstrip()
benchmark['Effective School Leadership %'] = benchmark['Effective School Leadership %'].str.rstrip()
benchmark['Strong Family-Community Ties %'] = benchmark['Strong Family-Community Ties %'].str.rstrip()
benchmark['Trust %'] = benchmark['Trust %'].str.rstrip('%').astype('float') / 100.
```

```
In [628]: benchmark.head(2)
```

```
Out[628]:
```

	Student Attendance Rate	Percent of Students Chronically Absent	\
0	0.94	0.18	
1	0.92	0.30	

	Rigorous Instruction %	Collaborative Teachers %	Supportive Environment %	\
0	0.89	0.94	0.86	
1	0.96	0.96	0.97	

	Effective School Leadership %	Strong Family-Community Ties %	Trust %	\
0	0.91	0.85	0.94	
1	0.90	0.86	0.94	

	Average ELA Proficiency	Average Math Proficiency	Count of Offers
0	2.14	2.17	NaN
1	2.63	2.98	NaN

```
In [629]: benchmark.fillna(0)
```

```
Out[629]:
```

	Student Attendance Rate	Percent of Students Chronically Absent	\
0	0.94	0.18	
1	0.92	0.30	
2	0.94	0.20	
3	0.92	0.28	
4	0.93	0.23	
5	0.92	0.33	

6	0.95	0.13
7	0.91	0.36
8	0.93	0.27
9	0.92	0.27
10	0.98	0.02
11	0.91	0.37
12	0.84	0.58
13	0.90	0.35
14	0.95	0.11
15	0.92	0.26
16	0.93	0.23
17	0.93	0.23
18	0.94	0.15
19	0.92	0.27
20	0.97	0.05
21	0.97	0.03
22	0.96	0.07
23	0.95	0.18
24	0.98	0.06
25	0.95	0.13
26	0.96	0.05
27	0.96	0.10
28	0.96	0.13
29	0.96	0.07
...
1334	0.00	0.00
1335	0.94	0.19
1336	0.00	0.00
1337	0.00	0.00
1338	0.96	0.09
1339	0.93	0.26
1340	1.00	0.00
1341	0.97	0.04
1342	0.95	0.11
1343	0.95	0.12
1344	0.00	0.00
1345	0.96	0.05
1346	0.97	0.06
1347	0.93	0.23
1348	0.95	0.14
1349	0.95	0.13
1350	0.94	0.20
1351	0.00	0.00
1352	0.00	1.00
1353	0.00	1.00
1354	0.95	0.33
1355	0.96	0.10
1356	0.91	0.34

1357	0.96	0.10
1358	0.96	0.10
1359	0.95	0.13
1360	0.94	0.24
1361	0.95	0.12
1362	0.95	0.12
1363	0.93	0.22

	Rigorous Instruction %	Collaborative Teachers % \
0	0.89	0.94
1	0.96	0.96
2	0.87	0.77
3	0.85	0.78
4	0.90	0.88
5	0.93	0.99
6	0.88	0.78
7	0.87	0.89
8	0.94	0.91
9	0.92	0.89
10	0.90	0.81
11	1.00	1.00
12	0.72	0.77
13	0.84	0.78
14	0.90	0.93
15	0.92	0.96
16	0.97	0.97
17	0.99	0.97
18	0.96	0.99
19	0.79	0.87
20	0.92	0.95
21	0.81	0.73
22	0.90	0.89
23	0.89	0.91
24	0.93	0.86
25	0.97	0.93
26	0.87	0.90
27	0.93	0.93
28	0.97	0.90
29	0.97	0.94
...
1334	0.00	0.00
1335	0.99	0.97
1336	0.00	0.00
1337	0.00	0.00
1338	0.91	0.92
1339	0.89	0.90
1340	0.96	0.96
1341	0.93	0.92

1342	0.95	0.93
1343	0.93	0.93
1344	0.00	0.00
1345	0.77	0.77
1346	0.99	0.99
1347	0.85	0.89
1348	0.90	0.95
1349	0.80	0.83
1350	0.95	0.84
1351	0.00	0.00
1352	0.97	0.92
1353	0.99	0.95
1354	0.96	0.98
1355	0.95	0.95
1356	0.74	0.83
1357	0.78	0.76
1358	0.84	0.88
1359	0.94	0.93
1360	0.93	0.90
1361	0.97	0.92
1362	0.93	0.91
1363	0.87	0.84

	Supportive Environment %	Effective School Leadership % \
0	0.86	0.91
1	0.97	0.90
2	0.82	0.61
3	0.82	0.73
4	0.87	0.81
5	0.95	0.91
6	0.95	0.69
7	0.88	0.88
8	0.85	0.87
9	0.90	0.83
10	0.91	0.67
11	0.99	0.99
12	0.77	0.72
13	0.81	0.80
14	0.94	0.93
15	0.92	0.96
16	0.97	0.96
17	0.95	0.96
18	0.96	0.98
19	0.82	0.77
20	0.92	0.96
21	0.85	0.55
22	0.88	0.86
23	0.95	0.86

24	0.94	0.76
25	0.97	0.93
26	0.95	0.87
27	0.98	0.91
28	0.94	0.82
29	0.99	0.91
...
1334	0.00	0.00
1335	0.97	0.92
1336	0.00	0.00
1337	0.00	0.00
1338	0.78	0.86
1339	0.93	0.88
1340	0.94	0.96
1341	0.88	0.84
1342	0.93	0.90
1343	0.79	0.86
1344	0.00	0.00
1345	0.87	0.70
1346	0.87	0.94
1347	0.83	0.83
1348	0.89	0.88
1349	0.81	0.83
1350	0.95	0.80
1351	0.00	0.00
1352	0.98	0.84
1353	0.98	0.88
1354	0.87	0.91
1355	0.96	0.86
1356	0.73	0.77
1357	0.78	0.71
1358	0.84	0.85
1359	0.94	0.88
1360	0.88	0.88
1361	0.89	0.84
1362	0.96	0.89
1363	0.84	0.77

	Strong Family-Community Ties %	Trust %	Average ELA Proficiency \
0	0.85	0.94	2.14
1	0.86	0.94	2.63
2	0.80	0.79	2.39
3	0.89	0.88	2.48
4	0.89	0.93	2.38
5	0.88	0.97	2.29
6	0.87	0.78	2.80
7	0.79	0.94	2.28
8	0.83	0.93	2.21

9	0.89	0.95	2.16
10	0.83	0.85	3.24
11	0.92	0.99	2.17
12	0.76	0.87	1.96
13	0.74	0.87	2.29
14	0.97	0.96	2.86
15	0.86	0.96	2.26
16	0.93	0.94	2.89
17	0.95	0.99	2.55
18	0.96	0.98	3.10
19	0.76	0.88	2.34
20	0.88	0.95	2.82
21	0.81	0.74	3.83
22	0.83	0.94	2.92
23	0.87	0.94	2.59
24	0.83	0.89	2.75
25	0.96	0.95	3.09
26	0.92	0.90	3.40
27	0.93	0.95	3.12
28	0.92	0.89	3.04
29	0.96	0.96	3.39
...
1334	0.00	0.00	0.00
1335	0.76	0.96	2.99
1336	0.00	0.00	0.00
1337	0.00	0.00	0.00
1338	0.83	0.90	2.69
1339	0.89	0.93	2.57
1340	0.91	0.98	2.43
1341	0.85	0.96	2.21
1342	0.84	0.95	2.59
1343	0.80	0.87	2.55
1344	0.00	0.00	0.00
1345	0.85	0.91	2.30
1346	0.92	0.97	2.38
1347	0.81	0.90	2.23
1348	0.90	0.94	2.67
1349	0.75	0.89	2.44
1350	0.83	0.91	2.83
1351	0.00	0.00	0.00
1352	0.93	0.97	3.23
1353	0.94	0.97	3.16
1354	0.90	0.95	3.00
1355	0.86	0.98	3.03
1356	0.86	0.91	2.02
1357	0.78	0.83	2.39
1358	0.87	0.91	2.42
1359	0.83	0.94	2.48

1360	0.88	0.93	2.50
1361	0.86	0.94	2.77
1362	0.91	0.95	2.60
1363	0.85	0.84	2.74

	Average Math Proficiency	Count of Offers
0	2.17	0.0
1	2.98	0.0
2	2.54	0.0
3	2.47	5.0
4	2.54	0.0
5	2.48	0.0
6	3.20	0.0
7	2.73	0.0
8	2.27	5.0
9	2.31	0.0
10	3.63	23.0
11	2.32	5.0
12	1.83	0.0
13	2.00	5.0
14	3.20	0.0
15	2.20	5.0
16	2.99	0.0
17	2.68	0.0
18	3.08	0.0
19	2.48	5.0
20	2.90	5.0
21	4.03	91.0
22	3.01	14.0
23	3.14	0.0
24	3.24	0.0
25	3.41	0.0
26	3.71	0.0
27	3.35	0.0
28	3.40	0.0
29	3.65	0.0
...
1334	0.00	0.0
1335	3.36	5.0
1336	0.00	0.0
1337	0.00	0.0
1338	2.81	5.0
1339	3.04	0.0
1340	2.66	5.0
1341	2.29	0.0
1342	2.76	5.0
1343	2.83	5.0
1344	0.00	5.0

1345	2.46	5.0
1346	2.30	5.0
1347	2.48	5.0
1348	2.87	5.0
1349	2.52	5.0
1350	3.11	5.0
1351	0.00	5.0
1352	3.97	5.0
1353	3.91	6.0
1354	3.52	5.0
1355	3.18	5.0
1356	2.27	0.0
1357	2.50	5.0
1358	2.87	5.0
1359	2.60	0.0
1360	2.85	5.0
1361	3.09	5.0
1362	3.29	0.0
1363	3.19	0.0

[1364 rows x 11 columns]

```
In [630]: #Check for nulls if any
benchmark.isnull().any()
```

```
Out[630]: Student Attendance Rate      True
Percent of Students Chronically Absent  True
Rigorous Instruction %                  True
Collaborative Teachers %               True
Supportive Environment %                True
Effective School Leadership %           True
Strong Family-Community Ties %         True
Trust %                                True
Average ELA Proficiency                 True
Average Math Proficiency                True
Count of Offers                        True
dtype: bool
```

```
In [631]: #Replace nulls if any
benchmark = benchmark.fillna(method='ffill')
```

```
In [632]: X1 = benchmark[[
    'Average ELA Proficiency',
    'Average Math Proficiency']]
```

```
In [633]: from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
X1_scaled = scaler.fit_transform( X1 )
```



```
In [634]: %matplotlib inline
```

```
cluster_range = range( 3, 9 ) #range excludes upper bound

for n_clusters in cluster_range:
    fig, (ax1, ax2) = plt.subplots(1, 2)
    fig.set_size_inches(18, 7)

    # The 1st subplot is the silhouette plot
    # The silhouette coefficient can range from -1, 1
    ax1.set_xlim([-1, 1])

    # The (n_clusters+1)*10 is for inserting blank space between silhouette
    # plots of individual clusters, to demarcate them clearly.
    ax1.set_ylim([0, len(X1_scaled) + (n_clusters + 1) * 10])

    cluster_labels = clusterer.fit_predict( X1_scaled )

    # The silhouette_score gives the average value for all the samples.
    # This gives a perspective into the density and separation of the formed
    # clusters
    silhouette_avg = silhouette_score(X1_scaled, cluster_labels)
    print("For n_clusters =", n_clusters,
          "The average silhouette_score is :", silhouette_avg)

    # Compute the silhouette scores for each sample
    sample_silhouette_values = silhouette_samples(X1_scaled, cluster_labels)

    y_lower = 10
    for i in range(n_clusters):
        # Aggregate the silhouette scores for samples belonging to
        # cluster i, and sort them
        ith_cluster_silhouette_values = \
            sample_silhouette_values[cluster_labels == i]

        ith_cluster_silhouette_values.sort()

        size_cluster_i = ith_cluster_silhouette_values.shape[0]
        y_upper = y_lower + size_cluster_i

        cmap = cm.get_cmap("Spectral")
        color = cmap(float(i) / n_clusters)

        #color = cm.spectral(float(i) / n_clusters)
        ax1.fill_betweenx(np.arange(y_lower, y_upper),
                          0, ith_cluster_silhouette_values,
                          facecolor=color, edgecolor=color, alpha=0.7)
```

```

        # Label the silhouette plots with their cluster numbers at the middle
        ax1.text(-0.05, y_lower + 0.5 * size_cluster_i, str(i))

        # Compute the new y_lower for next plot
        y_lower = y_upper + 10
        ax1.set_title("The silhouette plot for the various clusters.")
        ax1.set_xlabel("The silhouette coefficient values")
        ax1.set_ylabel("Cluster label")

        # The vertical line for average silhouette score of all the values
        ax1.axvline(x=silhouette_avg, color="red", linestyle="--")

        ax1.set_yticks([]) # Clear the yaxis labels / ticks
        ax1.set_xticks([-0.1, 0, 0.2, 0.4, 0.6, 0.8, 1])

        # 2nd Plot showing the actual clusters formed
        #cmap = cm.get_cmap("Spectral")
        colors = cmap(cluster_labels.astype(float) / n_clusters)
        #colors = cm.spectral(cluster_labels.astype(float) / n_clusters)
        ax2.scatter(X_scaled[:, 0], X_scaled[:, 1], marker='.', s=30, lw=0, alpha=0.7,
                    c=colors)

        # Labeling the clusters
        centers = kmeans.cluster_centers_
        #centers = clusterer.cluster_centers_
        # Draw white circles at cluster centers
        ax2.scatter(centers[:, 0], centers[:, 1],
                    marker='o', c="white", alpha=1, s=200)

        for i, c in enumerate(centers):
            ax2.scatter(c[0], c[1], marker='$%d$' % i, alpha=1, s=50)

        ax2.set_title("The visualization of the clustered data.")
        ax2.set_xlabel("Feature space for the 1st feature")
        ax2.set_ylabel("Feature space for the 2nd feature")

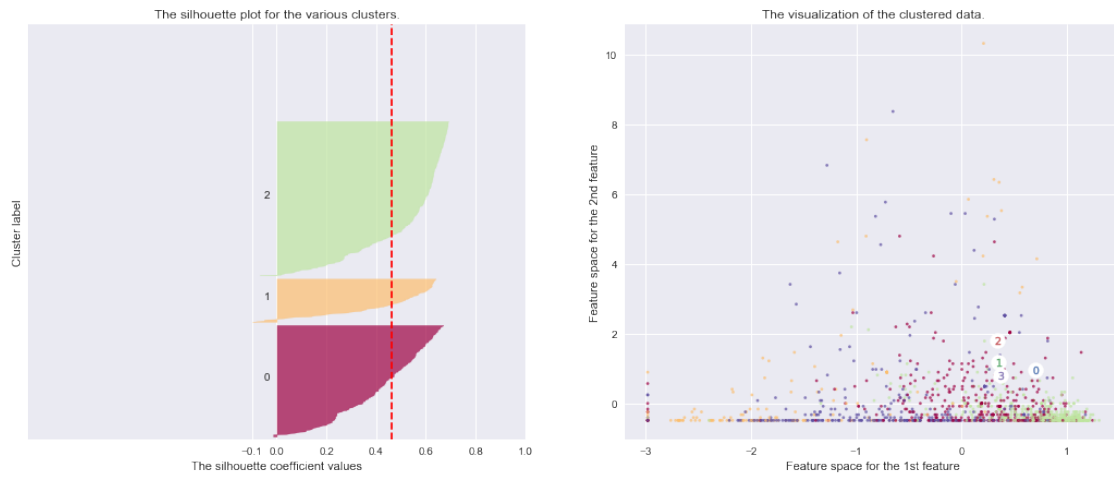
        plt.suptitle(("Silhouette analysis for KMeans clustering on sample data "
                    "with n_clusters = %d" % n_clusters),
                    fontsize=14, fontweight='bold')

        plt.show();

```

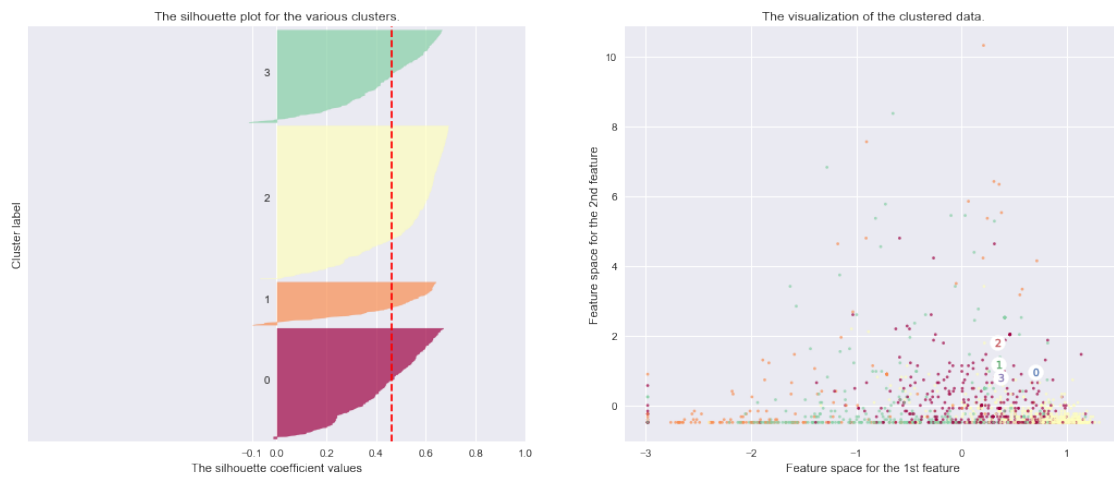
For n_clusters = 3 The average silhouette_score is : 0.462324434771

Silhouette analysis for KMeans clustering on sample data with n_clusters = 3

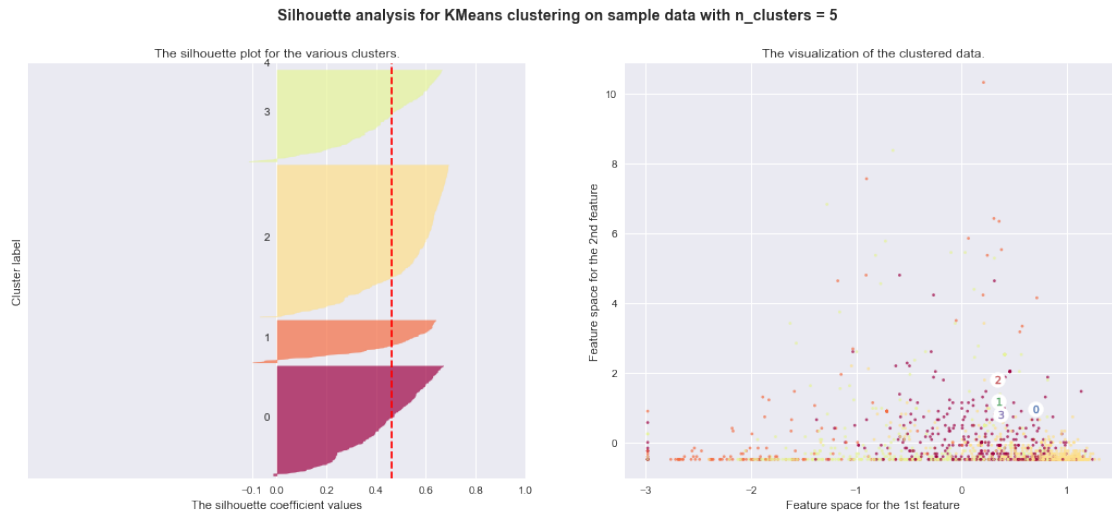


For n_clusters = 4 The average silhouette_score is : 0.462324434771

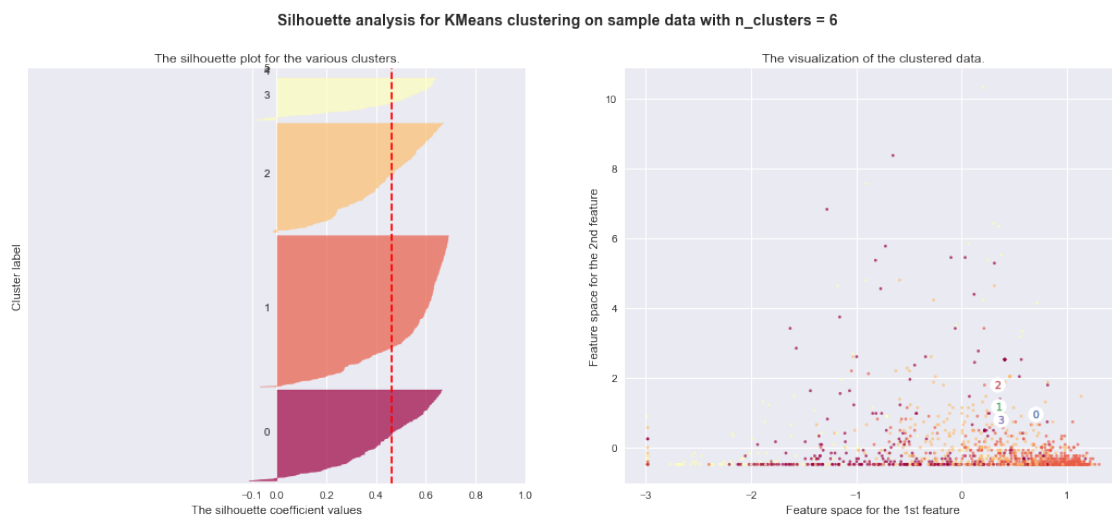
Silhouette analysis for KMeans clustering on sample data with n_clusters = 4



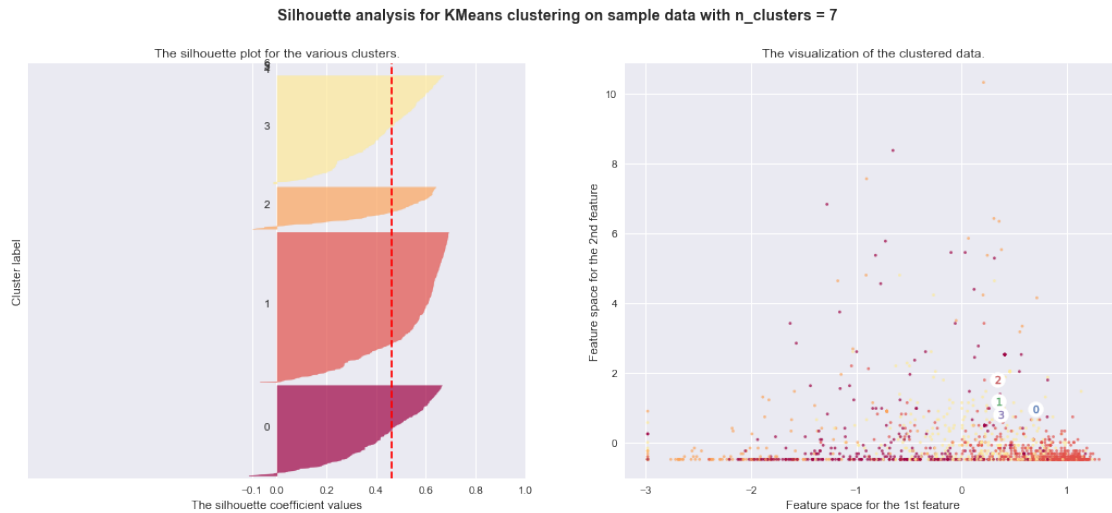
For n_clusters = 5 The average silhouette_score is : 0.462324434771



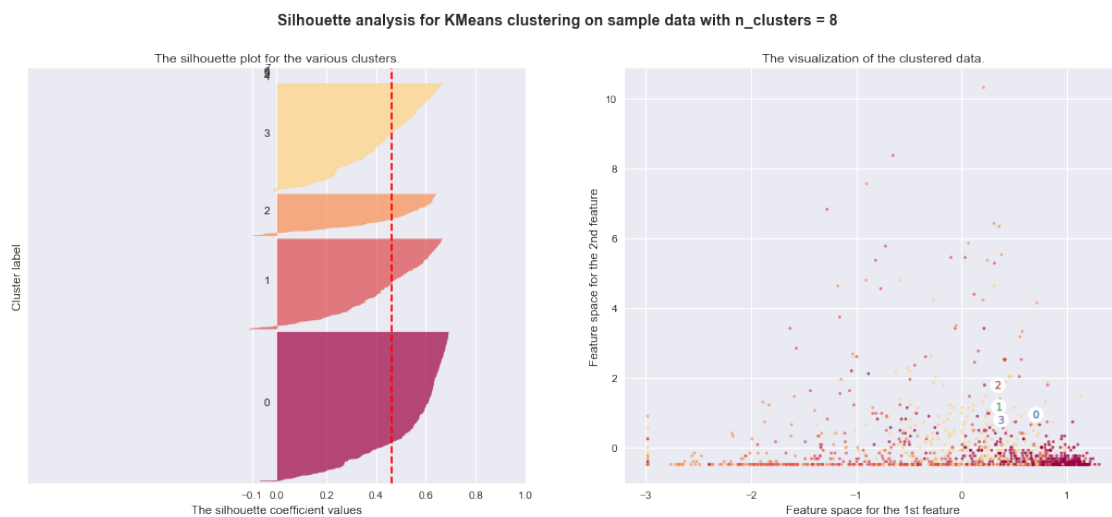
For n_clusters = 6 The average silhouette_score is : 0.462324434771



For n_clusters = 7 The average silhouette_score is : 0.462324434771



For n_clusters = 8 The average silhouette_score is : 0.462324434771



1.4 Results

- Section ??: Evaluation of Analysis
- Section ??: Validation based final inference

Section **1.5.3**: Back to Top

1.4.1 Model Evaluation and Validation

Silhouette score is the best for clusters =3 in our model using GMM technique. For K-means it shows best as clusters =4, and close to 0 for clusters =3, but the visuals clearly show that this is a result of overlapping. Given that Kmeans suffer from spherical clustering, this seems to validate the score as 3. When we compare with our benchmark model too, we get the silhouette score as same for clusters =3 to clusters =8 but the visuals clearly show the homogenous clustering for clusters =3. Thus, we can ascertain that clusters =3 is the optimal solution and comparable to benchmark model.

1.4.2 Justification

For the benchmark model as well, the best scores and homogenous clusters are for clusters = 3. It also is important to note that the original benchmark solution had ascertained cluster number as 3 though not using the techniques we employ. Thus, we can see that the clustering algorithms applied using our methods provides an equal measure of accuracy and hence, we can consider our model to be a good model.

1.5 Conclusion

Section 1.5.3: Back to Top

1.5.1 Reflection

In this project, the intent was to categorize the schools based on the variables, finding the optimal number of segments along the way. We did this using Kmeans and GMM with silhouette score as the basis of evaluation and comparison against benchmark model. One aspect that was interesting and challenging was determining the best silhouette score since we had very close values for the benchmark model for clusters numbering from 3 to 8. However, it is the visuals that help to pinpoint the right size. Again, for our model, GMM matched but not K-means; where it seemed like the right size did not match since 4 seemed to have best silhouette score value. Here too, the visuals provided a better comprehension with the overlapping effect evident (score of 0 again indicates close clusters which holds true with spherical constraints of k-means).

1.5.2 Improvement

One aspect of the analysis that can be improved is to use the analysis is to evaluate the performance using different seed values. Seed values were used in kmeans clustering and feature relevance prediction.

1.5.3 References

Please find below referenced articles and links in addition to those added in relevant sections.

- <http://www.awesomestats.in/python-cluster-validation/>
- https://scikit-learn.org/stable/auto_examples/cluster/plot_kmeans_silhouette_analysis.html#sphx-gl-r-auto-examples-cluster-plot-kmeans-silhouette-analysis-py
- <http://scikit-learn.org/stable/modules/clustering.html>
- <http://scikit-learn.org/stable/modules/generated/sklearn.metrics>
- [https://en.wikipedia.org/wiki/Elbow_method_\(clustering\)](https://en.wikipedia.org/wiki/Elbow_method_(clustering))

Section 1.5.3