



Education Disparity in Chicago Public High Schools

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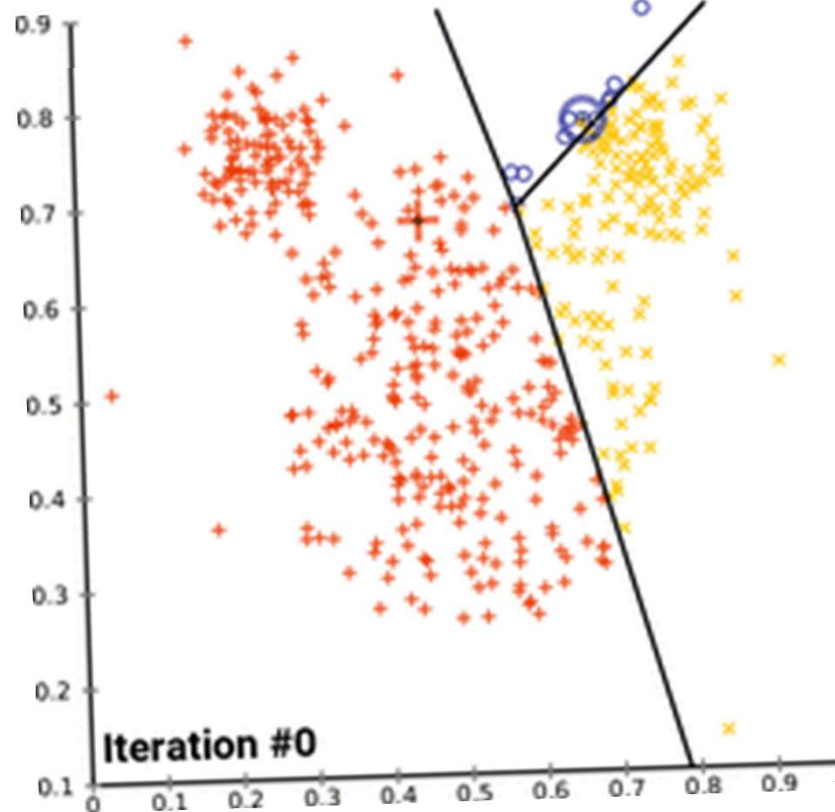




Clustering Schools

- I began by clustering schools based on data from ISBE using a K-Means clustering algorithm
- The K-Means algorithm took demographics as inputs like % Low Income, % White, % Black/African American, % Hispanic, etc.





- The goal of the algorithm is to have the “K means” converge to a “stable” point.
- The clusters are the partitions created by determining which mean each point is closest to.
- In this example, the black dots are the means and $K=3$

K-Means Clustering Algorithm



Summary Statistics (Cluster 0)

% Student Enrollment - White
1.11

% Student Enrollment - Black or African American
89.20

% Student Enrollment - Hispanic or Latino
8.38

% Student Enrollment - Asian
0.45

% Student Enrollment - Low Income
84.46

Student Attendance Rate
91.07

High School Dropout Rate - Total
5.44

High School 4-Year Graduation Rate - Total
80.92

% Graduates enrolled in a Postsecondary
Institution within 12 months 57.32



Summary Statistics (Cluster 1)

% Student Enrollment - White
4.15

% Student Enrollment - Black or African American
11.15

% Student Enrollment - Hispanic or Latino
81.30

% Student Enrollment - Asian
2.17

% Student Enrollment - Low Income
89.21

Student Attendance Rate
91.74

High School Dropout Rate - Total
4.41

High School 4-Year Graduation Rate - Total
81.06

% Graduates enrolled in a Postsecondary
Institution within 12 months 63.23



Summary Statistics (Cluster 2)

% Student Enrollment - White
28.25

% Student Enrollment - Black or African American
17.06

% Student Enrollment - Hispanic or Latino
38.19

% Student Enrollment - Asian
11.95

% Student Enrollment - Low Income
54.30

Student Attendance Rate
94.63

High School Dropout Rate - Total
2.24

High School 4-Year Graduation Rate - Total
91.25

% Graduates enrolled in a Postsecondary
Institution within 12 months 81.36





Trends over Time

- In addition to identifying summary statistics of the clusters, we can see how these clusters change over time
- More specifically, we are looking to analyze the demographic changes as well as academic changes



Graphical Model

2019 School Clusters

2020 School Clusters

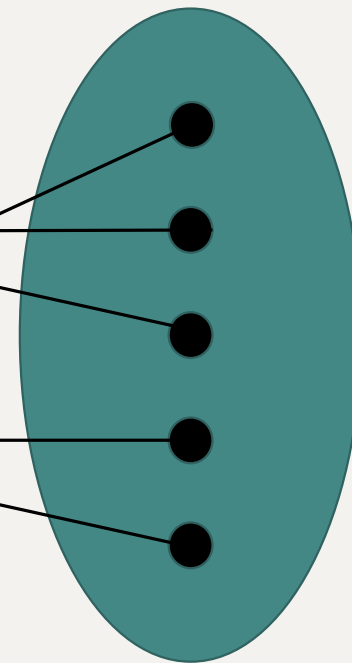
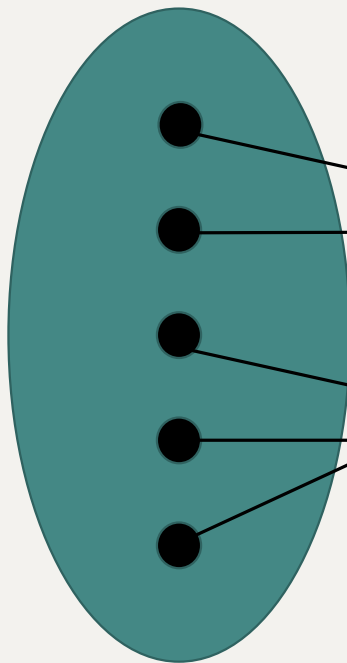
Cluster 1

Cluster 2

Cluster 3

Cluster 4

Cluster 5



Cluster 1

Cluster 2

Cluster 3

Cluster 4

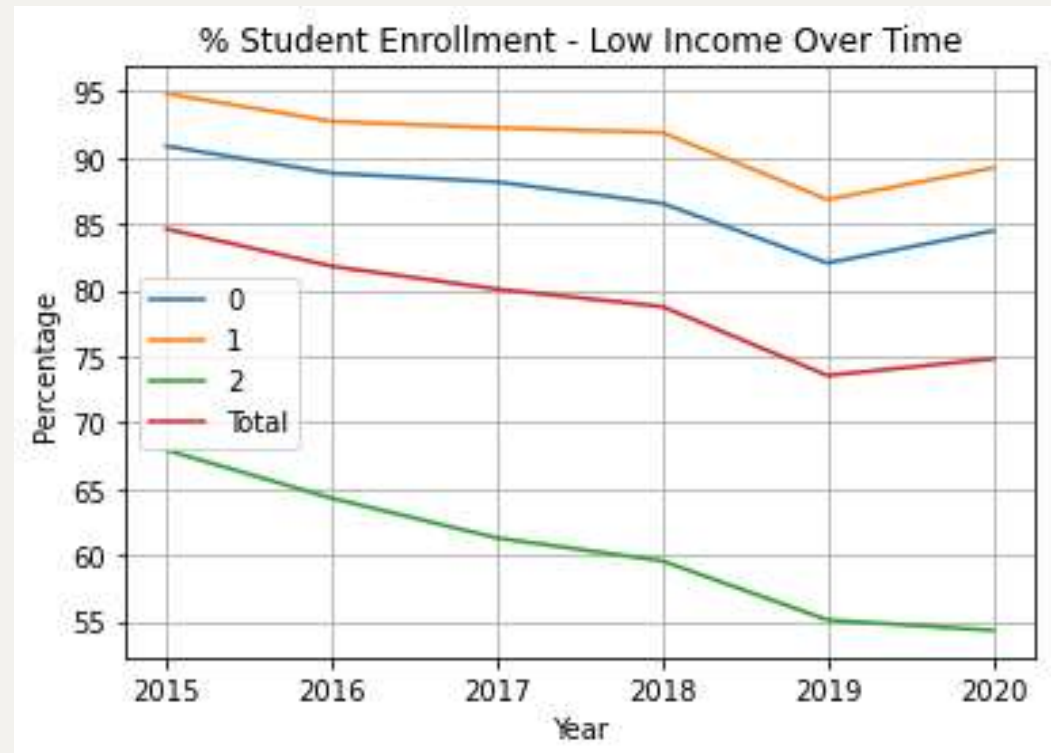
Cluster 5

Create stable matching between years



Low Income Proportions Over Time

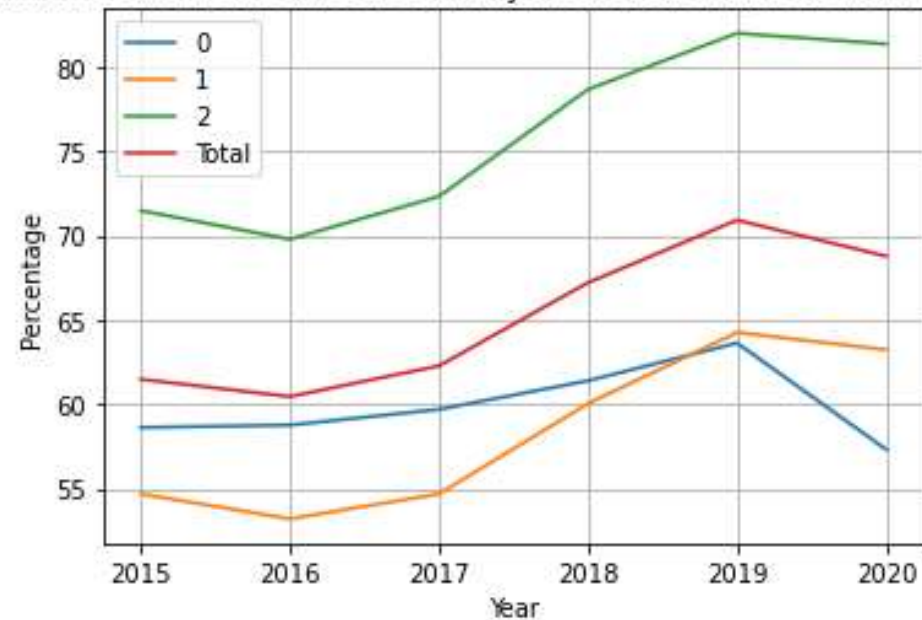
- As we can see, the gap between low-income student enrollment has gotten larger between the clusters.
- More specifically, this gap is mainly attributed to the decline in cluster 2.

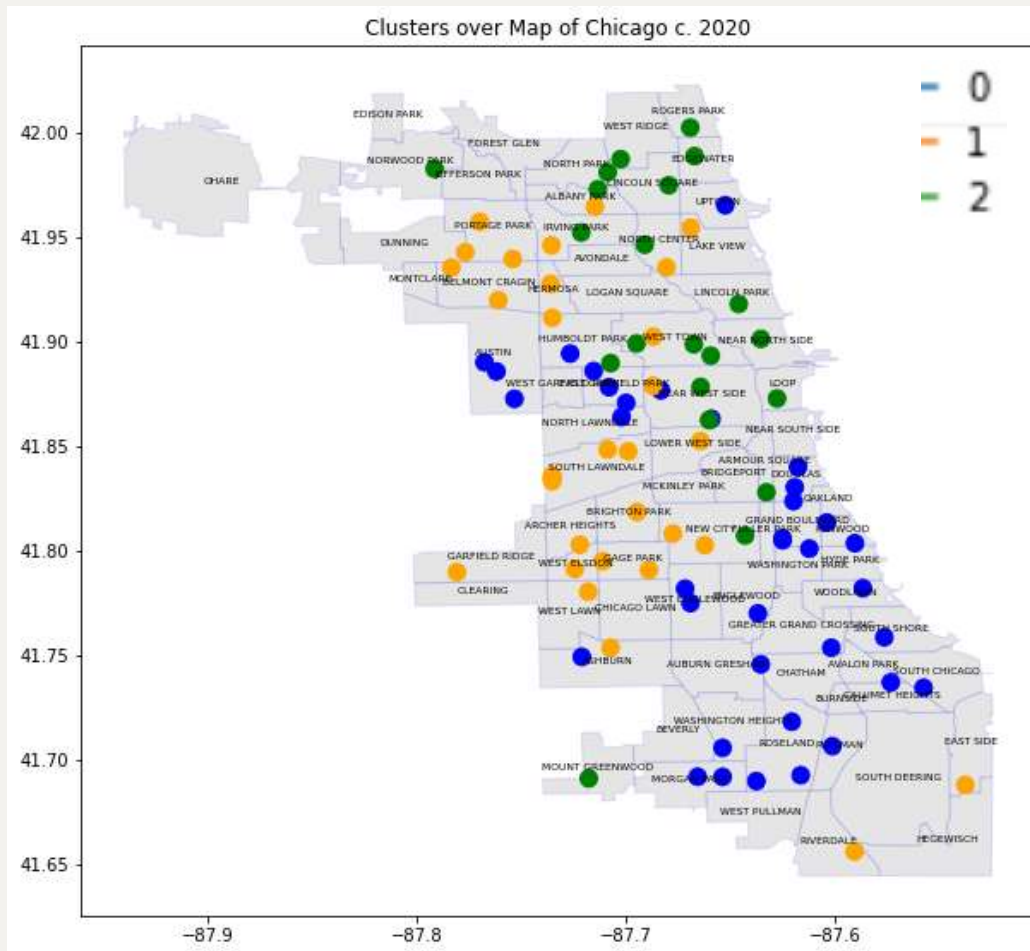


College Enrollment Over Time

- Again, the gap between college enrollment statistics has become even larger.
- More importantly, we can see how the pandemic halted growth rate of CPS students enrolling in college.
- We can also see how the pandemic affected students of color.

% Graduates enrolled in a Postsecondary Institution within 12 months Over Time





Mapping these Clusters

- After seeing these clusters projected onto a map of Chicago, we can see just how segregated the city is.
- Not only do the clusters give good indication of academic factors, but also partitions Chicago well.



Traditional Approach: Correlation Matrix

Index	% Student Enrollment - White	% Student Enrollment - Black or African American	% Student Enrollment - Hispanic or Latin American	% Student Enrollment - Asian	% Student Enrollment - Low Income	Student Attendance Rate	High School Dropout Rate - Total	High School 4-Year Graduation Rate - Total	% Graduates enrolled in a Postsecondary Institution
% Student Enrollment - White	1	-0.459694	0.0413344	0.670453	-0.881595	0.485443	-0.31436	0.313289	0.507657
% Student Enrollment - Black or African American	-0.459694	1	-0.897257	-0.420106	0.341374	-0.430198	0.269335	-0.232523	-0.432968
% Student Enrollment - Hispanic or Latin American	0.0413344	-0.897257	1	0.0520113	0.0392081	0.236293	-0.145118	0.101725	0.22719
% Student Enrollment - Asian	0.670453	-0.420106	0.0520113	1	-0.569967	0.413457	-0.250182	0.25526	0.434239
% Student Enrollment - Low Income	-0.881595	0.341374	0.0392081	-0.569967	1	-0.642394	0.447039	-0.483383	-0.644065
Student Attendance Rate	0.485443	-0.430198	0.236293	0.413457	-0.642394	1	-0.722903	0.784383	0.726142
High School Dropout Rate - Total	-0.31436	0.269335	-0.145118	-0.250182	0.447039	-0.722903	1	-0.741412	-0.547078
High School 4-Year Graduation Rate - Total	0.313289	-0.232523	0.101725	0.25526	-0.483383	0.784383	-0.741412	1	0.771031
% Graduates enrolled in a Postsecondary Institution	0.507657	-0.432968	0.22719	0.434239	-0.644065	0.726142	-0.547078	0.771031	1



Index	Student Enrollment - White	Student Enrollment - Black or African American	Student Enrollment - Hispanic or Latino	Student Enrollment - Asian or Pacific Islander	Student Enrollment - Low Income
Student Attendance Rate	0.485443	-0.430198	0.236293	0.413457	-0.642394
High School Dropout Rate - Total	-0.31436	0.269335	-0.145118	-0.250182	0.447039
High School 4-Year Graduation Rate - Total	0.313289	-0.232523	0.101725	0.25526	-0.483383
% Graduates enrolled in a Postsecondary Institution	0.507657	-0.432968	0.22719	0.434239	-0.644065

Inference

- From this we can see that in every academic factor, proportion of low-income and African American or Black students is the deciding factor.



Possible future work

Using other methods to cluster?

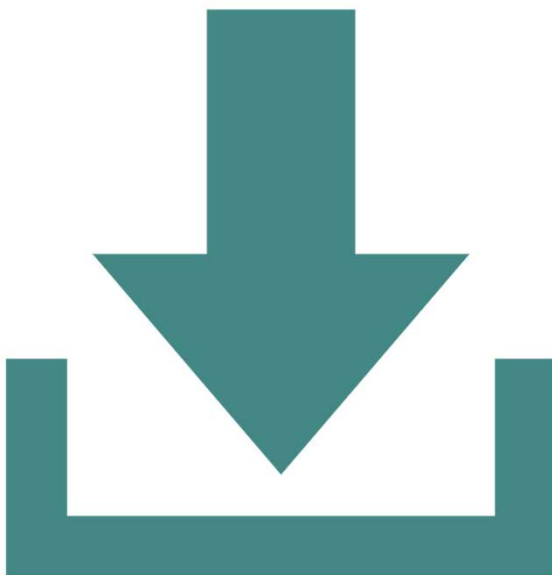
Analyzing different features?

Look into funding of schools?

Using methods in algebraic statistics

- Compute Markov basis functions from elimination Gröbner basis
- Construct Markov chain to “walk” along fibers
- Test for independence





Github Repo!

Progress can be found in my github repository
here: <https://github.com/mkralis123/SoReMo>

