

## Matthew Afsahi

In this project, I will be using Apache Spark engine and SQL to queries the data sets.

```
1 from google.colab import drive
2 drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount(force=True)

```
1 # Installing java, apache spark and related libraries
2
3 !apt-get install openjdk-8-jdk-headless -qq > /dev/null
4
5 # install spark (change the version number if needed)
6 !wget -q https://archive.apache.org/dist/spark/spark-3.0.0/spark-3.0.0-bin-hadoop3.tgz
7
8 # unzip the spark file to the current folder
9 !tar xf spark-3.0.0-bin-hadoop3.2.tgz
10
11 # set your spark folder to your system path environment.
12 import os
13 os.environ["JAVA_HOME"] = "/usr/lib/jvm/java-8-openjdk-amd64"
14 os.environ["SPARK_HOME"] = "/content/spark-3.0.0-bin-hadoop3.2"
15
16
17 # install findspark using pip
18 !pip install -q findspark
19
20 !pip install pyspark
21
22
```

Requirement already satisfied: pyspark in /usr/local/lib/python3.6/dist-packages (3.0.0)  
Requirement already satisfied: py4j==0.10.9 in /usr/local/lib/python3.6/dist-packages (0.10.9)

```
1 # Initiating the spark on this notebook
2 import findspark
```

```
1 findspark.init('/content/spark-3.0.0-bin-hadoop3.2')
```

```

1 from pyspark.sql import SparkSession

2

1 spark=SparkSession.builder.appName('MyAssignment_3').getOrCreate()

2

1 #Reading the census data set from the cloud
2 census=spark.read.csv('/content/drive/My Drive/Colab Notebooks/data/Ce

1 census.printSchema()

root
|-- COMMUNITY_AREA_NUMBER: integer (nullable = true)
|-- COMMUNITY_AREA_NAME: string (nullable = true)
|-- PERCENT OF HOUSING CROWDED: double (nullable = true)
|-- PERCENT HOUSEHOLDS BELOW POVERTY: double (nullable = true)
|-- PERCENT AGED 16+ UNEMPLOYED: double (nullable = true)
|-- PERCENT AGED 25+ WITHOUT HIGH SCHOOL DIPLOMA: double (nullable = true)
|-- PERCENT AGED UNDER 18 OR OVER 64: double (nullable = true)
|-- PER_CAPITA_INCOME : integer (nullable = true)
|-- HARDSHIP_INDEX: integer (nullable = true)

1 #Reading the school data set from the cloud
2 school=spark.read.csv('/content/drive/My Drive/Colab Notebooks/data/C

1 school.printSchema()

|-- Environment Icon : string (nullable = true)
|-- Environment Score: integer (nullable = true)

|-- Instruction Icon : string (nullable = true)
|-- Instruction Score: integer (nullable = true)
|-- Leaders Icon : string (nullable = true)
|-- Leaders Score : string (nullable = true)
|-- Teachers Icon : string (nullable = true)
|-- Teachers Score: string (nullable = true)
|-- Parent Engagement Icon : string (nullable = true)
|-- Parent Engagement Score: string (nullable = true)
|-- Parent Environment Icon: string (nullable = true)
|-- Parent Environment Score: string (nullable = true)
|-- AVERAGE_STUDENT_ATTENDANCE: string (nullable = true)
|-- Rate of Misconducts (per 100 students) : double (nullable = true)
|-- Average Teacher Attendance: string (nullable = true)
|-- Individualized Education Program Compliance Rate : string (nullable = true)
|-- Pk-2 Literacy %: string (nullable = true)
|-- Pk-2 Math %: string (nullable = true)
|-- Gr3-5 Grade Level Math %: string (nullable = true)
|-- Gr3-5 Grade Level Read % : string (nullable = true)
|-- Gr3-5 Keep Pace Read %: string (nullable = true)
|-- Gr3-5 Keep Pace Math %: string (nullable = true)
|-- Gr6-8 Grade Level Math %: string (nullable = true)
|-- Gr6-8 Grade Level Read %: string (nullable = true)

```

```

-- Gr6-8 Keep Pace Math%: string (nullable = true)
-- Gr6-8 Keep Pace Read %: string (nullable = true)
-- Gr-8 Explore Math %: string (nullable = true)
-- Gr-8 Explore Read %: string (nullable = true)
-- ISAT Exceeding Math %: double (nullable = true)
-- ISAT Exceeding Reading % : double (nullable = true)
-- ISAT Value Add Math: double (nullable = true)
-- ISAT Value Add Read: double (nullable = true)
-- ISAT Value Add Color Math: string (nullable = true)
-- ISAT Value Add Color Read: string (nullable = true)
-- Students Taking Algebra %: string (nullable = true)
-- Students Passing Algebra %: string (nullable = true)
-- 9th Grade EXPLORE (2009) : string (nullable = true)
-- 9th Grade EXPLORE (2010) : string (nullable = true)
-- 10th Grade PLAN (2009) : string (nullable = true)
-- 10th Grade PLAN (2010) : string (nullable = true)
-- Net Change EXPLORE and PLAN: string (nullable = true)
-- 11th Grade Average ACT (2011) : string (nullable = true)
-- Net Change PLAN and ACT: string (nullable = true)
-- College Eligibility %: string (nullable = true)
-- Graduation Rate %: string (nullable = true)
-- College Enrollment Rate %: string (nullable = true)
-- COLLEGE_ENROLLMENT: integer (nullable = true)
-- General Services Route : integer (nullable = true)
-- Freshman on Track Rate %: string (nullable = true)
-- X_COORDINATE: double (nullable = true)
-- Y_COORDINATE: double (nullable = true)
-- Latitude: double (nullable = true)
-- Longitude: double (nullable = true)
-- COMMUNITY_AREA_NUMBER: integer (nullable = true)
-- COMMUNITY_AREA_NAME: string (nullable = true)
-- Ward: integer (nullable = true)
-- Police District: integer (nullable = true)
-- Location: string (nullable = true)

```

1 census.show(10)

```

+-----+-----+-----+-----+
|COMMUNITY_AREA_NUMBER|COMMUNITY_AREA_NAME|PERCENT OF HOUSING CROWDED|PERCENT HOUSEH|
+-----+-----+-----+-----+
|          1|      Rogers Park|          7.7|
|          2|      West Ridge|          7.8|
|          3|          Uptown|          3.8|
|          4|    Lincoln Square|          3.4|
|          5|      North Center|          0.3|
|          6|      Lake View|          1.1|
|          7|      Lincoln Park|          0.8|
|          8|    Near North Side|          1.9|
|          9|      Edison Park|          1.1|
|         10|      Norwood Park|          2.0|
+-----+-----+-----+-----+
only showing top 10 rows

```

1 # Creating a SQL temp for more queries, Note that SQL is working on tc  
2 census.createOrReplaceTempView('Census')

Q1. What was the per capita income in North Park community?

26576

```
1 spark.sql("select * \
2         from Census\
3         where Census.COMMUNITY_AREA_NAME== 'North Park'").show()
```

COMMUNITY_AREA_NUMBER	COMMUNITY_AREA_NAME	PERCENT OF HOUSING CROWDED	PERCENT HOUSEHOLDS
13	North Park	3.9	

Q2. How many community area were in the census data base? 77

```
1 spark.sql("select count(COMMUNITY_AREA_NUMBER) \
2         from Census ").show()
3
```

count(COMMUNITY_AREA_NUMBER)
77

```
1 school.show(10)
```

School ID	NAME_OF_SCHOOL	Elementary, Middle, or High School	Street Address
610038	Abraham Lincoln E...	ES	615 W Kemper F
610281	Adam Clayton Powe...	ES	7511 S South Shor
610185	Adlai E Stevenson...	ES	8010 S Kostner Av
609993	Agustin Lara Elem...	ES	4619 S Wolcott Av
610513	Air Force Academy...	HS	3630 S Wells S
610212	Albany Park Multi...	MS	4929 N Sawyer Av
609720	Albert G Lane Tec...	HS	2501 W Addison S
610342	Albert R Sabin El...	ES	2216 W Hirsch S
610524	Alcott High Schoo...	HS	2957 N Hoyne Av
610209	Alessandro Volta ...	ES	4950 N Avers Av

only showing top 10 rows

```
1 # Creating the School temp data base on the cloud
```

```
1 # creating the school temp data base on the cloud
2 school.createOrReplaceTempView('School')
```

Q3. How many unique School are in Chicago data base system?

566

```
1 spark.sql("select distinct count(NAME_OF_SCHOOL) \
2          from School").show()
```

```
+-----+
|count(NAME_OF_SCHOOL)|
+-----+
|                    566|
+-----+
```

```
1 census.describe( ).show()
```

```
+-----+-----+-----+-----+-----+
|summary|COMMUNITY_AREA_NUMBER|COMMUNITY_AREA_NAME|PERCENT OF HOUSING CROWDED|PERCENT|
+-----+-----+-----+-----+-----+
| count|                77|                78|                78|
|  mean|                39.0|                null|4.920512820512823|
| stddev|22.371857321197094|                null|3.6589814413502|
|   min|                  1|Albany Park|0.3|
|   max|                77|Woodlawn|15.8|
+-----+-----+-----+-----+-----+
```

```
1 census.summary().show()
```

```
+-----+-----+-----+-----+-----+
|summary|COMMUNITY_AREA_NUMBER|COMMUNITY_AREA_NAME|PERCENT OF HOUSING CROWDED|PERCENT|
+-----+-----+-----+-----+-----+
| count|                77|                78|                78|
|  mean|                39.0|                null|4.920512820512823|
| stddev|22.371857321197094|                null|3.6589814413502|
|   min|                  1|Albany Park|0.3|
|  25%|                20|                null|2.3|
|  50%|                39|                null|3.8|
|  75%|                58|                null|6.8|
|   max|                77|Woodlawn|15.8|
+-----+-----+-----+-----+-----+
```

```
1 census.toPandas().shape
```

```
(78, 9)
```

```
1 census.toPandas().isnull().sum()
```

```

COMMUNITY_AREA_NUMBER      1
COMMUNITY_AREA_NAME         0
PERCENT OF HOUSING CROWDED  0
PERCENT HOUSEHOLDS BELOW POVERTY  0
PERCENT AGED 16+ UNEMPLOYED  0
PERCENT AGED 25+ WITHOUT HIGH SCHOOL DIPLOMA  0
PERCENT AGED UNDER 18 OR OVER 64  0
PER_CAPITA_INCOME           0
HARDSHIP_INDEX              1
dtype: int64

```

```
1 census.createOrReplaceTempView('censusDataSQL')
```

Q4. What records or record has found with no community area

- number in census data base? No name and No community area name has found.

```

1 spark.sql('select * \
2           from censusDataSQL \
3           where COMMUNITY_AREA_NUMBER IS NULL').show()

```

```

+-----+-----+-----+-----+
|COMMUNITY_AREA_NUMBER|COMMUNITY_AREA_NAME|PERCENT OF HOUSING CROWDED|PERCENT HOUSEHOLDS BELOW POVERTY|
+-----+-----+-----+-----+
|              null|      CHICAGO|              4.7|
+-----+-----+-----+-----+

```

```

1 spark.sql('select * \
2           from censusDataSQL \
3           where HARDSHIP_INDEX IS NULL').show()

```

```

+-----+-----+-----+-----+
|COMMUNITY_AREA_NUMBER|COMMUNITY_AREA_NAME|PERCENT OF HOUSING CROWDED|PERCENT HOUSEHOLDS BELOW POVERTY|
+-----+-----+-----+-----+
|              null|      CHICAGO|              4.7|
+-----+-----+-----+-----+

```

```
1 school.show(5)
```

```

+-----+-----+-----+-----+
|School ID|NAME_OF_SCHOOL|Elementary, Middle, or High School|Street Address|
+-----+-----+-----+-----+
|  610038|Abraham Lincoln E...|ES|615 W Kemper F

```

```
| 610281|Adam Clayton Powe...| ES|7511 S South Shor
| 610185|Adlai E Stevenson...| ES| 8010 S Kostner Av
| 609993|Agustin Lara Elem...| ES| 4619 S Wolcott Av
| 610513|Air Force Academy...| HS| 3630 S Wells S
+-----+-----+-----+-----+
only showing top 5 rows
```

```
1 # Dropping the row wich will not affect the other rows.
2
3 census=census.na.drop()
4
```

```
1 # checking wich records have the null values?
2 census.toPandas().isnull().sum()
```

```
COMMUNITY_AREA_NUMBER      0
COMMUNITY_AREA_NAME         0
PERCENT OF HOUSING CROWDED  0
PERCENT HOUSEHOLDS BELOW POVERTY  0
PERCENT AGED 16+ UNEMPLOYED  0
PERCENT AGED 25+ WITHOUT HIGH SCHOOL DIPLOMA  0
PERCENT AGED UNDER 18 OR OVER 64  0
PER_CAPITA_INCOME           0
HARDSHIP_INDEX              0
dtype: int64
```

```
1 census.toPandas().shape

(77, 9)
```

```
1 # Checking the value data types in our data. In order to fir the data
2
3 census.toPandas().dtypes
```

```
COMMUNITY_AREA_NUMBER      int32
COMMUNITY_AREA_NAME        object
PERCENT OF HOUSING CROWDED  float64
PERCENT HOUSEHOLDS BELOW POVERTY  float64
PERCENT AGED 16+ UNEMPLOYED  float64
PERCENT AGED 25+ WITHOUT HIGH SCHOOL DIPLOMA  float64
PERCENT AGED UNDER 18 OR OVER 64  float64
PER_CAPITA_INCOME          int32
HARDSHIP_INDEX             int32
dtype: object
```

```
1 # Sine we have a unique number for the community area name we do not r
2
3 census=census.drop(census.COMMUNITY_AREA_NAME)
```

```
1 census.toPandas().dtypes
```

```
1 census.toPandas().dtypes
```

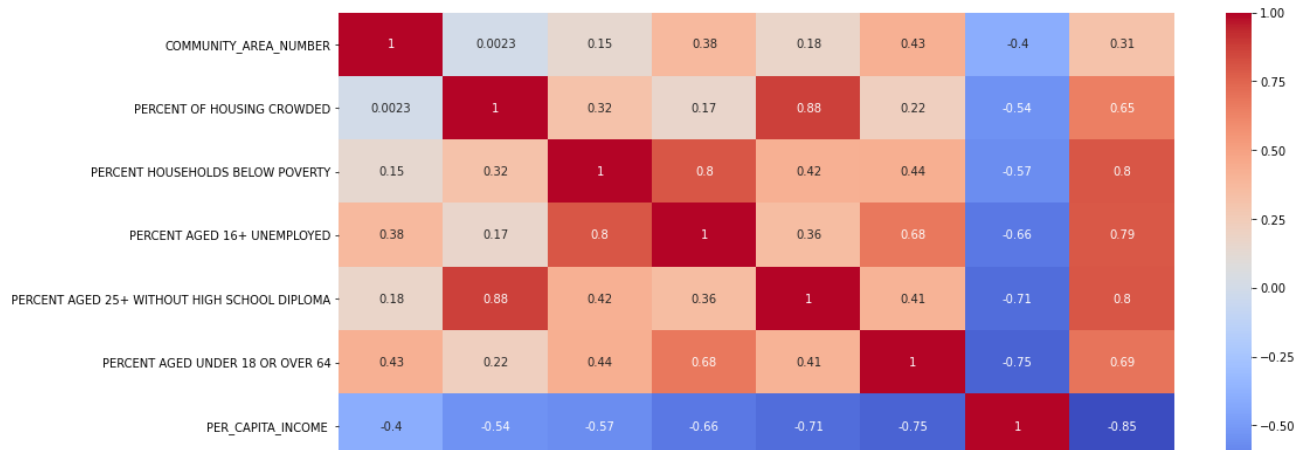
```
COMMUNITY_AREA_NUMBER      int32
PERCENT OF HOUSING CROWDED  float64
PERCENT HOUSEHOLDS BELOW POVERTY  float64
PERCENT AGED 16+ UNEMPLOYED  float64
PERCENT AGED 25+ WITHOUT HIGH SCHOOL DIPLOMA  float64
PERCENT AGED UNDER 18 OR OVER 64  float64
PER_CAPITA_INCOME          int32
HARDSHIP_INDEX             int32
dtype: object
```

With a quick glance we can have an idea that which variables may be more effective for the deep finding which variables have more stronger coefficient we need to feature extraction process, which I will impliment those methods later on this project.

```
1 import seaborn as sns
2 import matplotlib.pyplot as plt
3 % matplotlib inline
4 plt.figure(figsize=(16,8))
5 sns.heatmap(census.toPandas().corr(),annot=True,cmap='coolwarm')
```



<matplotlib.axes.\_subplots.AxesSubplot at 0x7f8bfdab0b70>



```
1 census.show(2)
```

```
+-----+-----+-----+-----+
|COMMUNITY_AREA_NUMBER|PERCENT OF HOUSING CROWDED|PERCENT HOUSEHOLDS BELOW POVERTY|PERCENT AGED 16+ UNEMPLOYED|
+-----+-----+-----+-----+
|                1|                7.7|                23.6|                7.7|
|                2|                7.8|                17.2|                7.8|
+-----+-----+-----+-----+
```

only showing top 2 rows

```
1 y=census.select('PER_CAPITA_INCOME ').collect()
```

```
1 X=census.drop('PER_CAPITA_INCOME ').collect()
```

```
1 census.columns
```

```
['COMMUNITY_AREA_NUMBER',
 'PERCENT OF HOUSING CROWDED',
 'PERCENT HOUSEHOLDS BELOW POVERTY',
 'PERCENT AGED 16+ UNEMPLOYED',
 'PERCENT AGED 25+ WITHOUT HIGH SCHOOL DIPLOMA',
 'PERCENT AGED UNDER 18 OR OVER 64',
 'PER_CAPITA_INCOME ',
 'HARDSHIP_INDEX']
```

Feature Selection and Extraction to find the most effective explanatory variable in our data. Here I am using  
 ExtraTreesClassifier to find out and later on I wil implimenting more effective methods to do so.

```
1 from sklearn.ensemble import ExtraTreesClassifier
```

```
1 model=ExtraTreesClassifier()
```

```
1 model.fit(X,y)
```

```
/usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:1: DataConversionWarning
    """Entry point for launching an IPython kernel.
```

```
ExtraTreesClassifier(bootstrap=False, ccp_alpha=0.0, class_weight=None,
    criterion='gini', max_depth=None, max_features='auto',
    max_leaf_nodes=None, max_samples=None,
    min_impurity_decrease=0.0, min_impurity_split=None,
    min_samples_leaf=1, min_samples_split=2,
    min_weight_fraction_leaf=0.0, n_estimators=100,
    n_jobs=None, oob_score=False, random_state=None, verbose=0,
    warm_start=False)
```

```
1 import pandas as pd
2 feat_importance=pd.Series(model.feature_importances_,index=[ 'COMMUNITY
3   'PERCENT OF HOUSING CROWDED',
4   'PERCENT HOUSEHOLDS BELOW POVERTY',
5   'PERCENT AGED 16+ UNEMPLOYED',
6   'PERCENT AGED 25+ WITHOUT HIGH SCHOOL DIPLOMA',
7   'PERCENT AGED UNDER 18 OR OVER 64',
8   'HARDSHIP_INDEX'])
```

```
1 feat_importance.nlargest(7).plot(kind='bar')
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f8bf4168d68>



in addition to the correlation plot I tested to which variables should

I keep and importance feature testing and it turns out that the most variables

are highly correlated each other. Therefore, I will keep the rest of the

variables in this data.

```

1 school.show(10)

+-----+-----+-----+-----+
|School ID|      NAME_OF_SCHOOL|Elementary, Middle, or High School|      Street Address|
+-----+-----+-----+-----+
|  610038|Abraham Lincoln E...|      ES|      615 W Kemper F
|  610281|Adam Clayton Powe...|      ES| 7511 S South Shor
|  610185|Adlai E Stevenson...|      ES| 8010 S Kostner Av
|  609993|Agustin Lara Elem...|      ES| 4619 S Wolcott Av
|  610513|Air Force Academy...|      HS|      3630 S Wells S
|  610212|Albany Park Multi...|      MS| 4929 N Sawyer Av
|  609720|Albert G Lane Tec...|      HS| 2501 W Addison S
|  610342|Albert R Sabin El...|      ES|      2216 W Hirsch S
|  610524|Alcott High Schoo...|      HS|      2957 N Hoyne Av
|  610209|Alessandro Volta ...|      ES|      4950 N Avers Av
+-----+-----+-----+-----+
only showing top 10 rows

```

```

1 # More SQL
2 school.createOrReplaceTempView('SchoolDataSQL')

```

```

1 school.toPandas().shape

(566, 78)

```

# Finding which variables in our data is object type, as we see there

## 1 school.dtypes

```
( 'Family Involvement Score', 'string'),
( 'Environment Icon ', 'string'),
( 'Environment Score', 'int'),
( 'Instruction Icon ', 'string'),
( 'Instruction Score', 'int'),
( 'Leaders Icon ', 'string'),
( 'Leaders Score ', 'string'),
( 'Teachers Icon ', 'string'),
( 'Teachers Score', 'string'),
( 'Parent Engagement Icon ', 'string'),
( 'Parent Engagement Score', 'string'),
( 'Parent Environment Icon', 'string'),
( 'Parent Environment Score', 'string'),
( 'AVERAGE_STUDENT_ATTENDANCE', 'string'),
( 'Rate of Misconducts (per 100 students) ', 'double'),
( 'Average Teacher Attendance', 'string'),
( 'Individualized Education Program Compliance Rate ', 'string'),
( 'Pk-2 Literacy %', 'string'),
( 'Pk-2 Math %', 'string'),
( 'Gr3-5 Grade Level Math %', 'string'),
( 'Gr3-5 Grade Level Read % ', 'string'),
( 'Gr3-5 Keep Pace Read %', 'string'),
( 'Gr3-5 Keep Pace Math %', 'string'),
( 'Gr6-8 Grade Level Math %', 'string'),
( 'Gr6-8 Grade Level Read %', 'string'),
( 'Gr6-8 Keep Pace Math%', 'string'),
( 'Gr6-8 Keep Pace Read %', 'string'),
( 'Gr-8 Explore Math %', 'string'),
( 'Gr-8 Explore Read %', 'string'),
( 'ISAT Exceeding Math %', 'double'),
( 'ISAT Exceeding Reading % ', 'double'),
( 'ISAT Value Add Math', 'double'),
( 'ISAT Value Add Read', 'double'),
( 'ISAT Value Add Color Math', 'string'),
( 'ISAT Value Add Color Read', 'string'),
( 'Students Taking Algebra %', 'string'),
( 'Students Passing Algebra %', 'string'),

( '9th Grade EXPLORE (2009) ', 'string'),
( '9th Grade EXPLORE (2010) ', 'string'),
( '10th Grade PLAN (2009) ', 'string'),
( '10th Grade PLAN (2010) ', 'string'),
( 'Net Change EXPLORE and PLAN', 'string'),
( '11th Grade Average ACT (2011) ', 'string'),
( 'Net Change PLAN and ACT', 'string'),
( 'College Eligibility %', 'string'),
( 'Graduation Rate %', 'string'),
( 'College Enrollment Rate %', 'string'),
( 'COLLEGE_ENROLLMENT', 'int'),
( 'General Services Route ', 'int'),
( 'Freshman on Track Rate %', 'string'),
( 'X_COORDINATE', 'double'),
( 'Y_COORDINATE', 'double'),
( 'Latitude', 'double'),
( 'Longitude', 'double'),
( 'COMMUNITY_AREA_NUMBER', 'int'),
( 'COMMUNITY_AREA_NAME', 'string'),
( 'Ward ', 'int')
```

```
( 'waru', int ),
('Police District', 'int'),
('Location', 'string')]
```

## ▼ Looping through the school data set to find the String Objects.

```
1 # the columns which are string and not numeric
2 c=0
3 for i,j in school.dtypes:
4     if j=='string':
5         c+=1
6         print(i)
7
8 print('-----')
9 print('\n')
10
11 print(f'There are {c} categorical variable in school data set')
12
```

```
State
Phone Number
Link
Network Manager
Collaborative Name
Adequate Yearly Progress Made?
Track Schedule
CPS Performance Policy Status
CPS Performance Policy Level
HEALTHY_SCHOOL_CERTIFIED
Safety Icon
Family Involvement Icon
Family Involvement Score
Environment Icon
Instruction Icon
Leaders Icon
Leaders Score
Teachers Icon
Teachers Score
Parent Engagement Icon
Parent Engagement Score
Parent Environment Icon
Parent Environment Score
AVERAGE_STUDENT_ATTENDANCE
Average Teacher Attendance
Individualized Education Program Compliance Rate
Pk-2 Literacy %
Pk-2 Math %
Gr3-5 Grade Level Math %
Gr3-5 Grade Level Read %
Gr3-5 Keep Pace Read %
Gr3-5 Keep Pace Math %
Gr6-8 Grade Level Math %
Gr6-8 Grade Level Read %
Gr6-8 Keep Pace Math%
Gr6-8 Keep Pace Read %
Gr 8 Evaluate Math %
```

```

Gr-8 EXPLORE Math %
Gr-8 EXPLORE Read %
ISAT Value Add Color Math
ISAT Value Add Color Read
Students Taking Algebra %
Students Passing Algebra %
9th Grade EXPLORE (2009)
9th Grade EXPLORE (2010)
10th Grade PLAN (2009)
10th Grade PLAN (2010)
Net Change EXPLORE and PLAN
11th Grade Average ACT (2011)
Net Change PLAN and ACT
College Eligibility %
Graduation Rate %
College Enrollment Rate %
Freshman on Track Rate %
COMMUNITY_AREA_NAME
Location
-----

```

```
1 school.show(10)
```

```

+-----+-----+-----+-----+
|School ID|NAME_OF_SCHOOL|Elementary, Middle, or High School|Street Address|
+-----+-----+-----+-----+
| 610038|Abraham Lincoln E...|ES|615 W Kemper F
| 610281|Adam Clayton Powe...|ES|7511 S South Shor
| 610185|Adlai E Stevenson...|ES|8010 S Kostner Av
| 609993|Agustin Lara Elem...|ES|4619 S Wolcott Av
| 610513|Air Force Academy...|HS|3630 S Wells S
| 610212|Albany Park Multi...|MS|4929 N Sawyer Av
| 609720|Albert G Lane Tec...|HS|2501 W Addison S
| 610342|Albert R Sabin El...|ES|2216 W Hirsch S
| 610524|Alcott High Schoo...|HS|2957 N Hoyne Av
| 610209|Alessandro Volta ...|ES|4950 N Avers Av
+-----+-----+-----+-----+
only showing top 10 rows

```

There are too many variable in school data sets that should be drop because many of them have the same informatio, for example since we know that this data set is for Chicago city and IL state, so we do not need to repeat same information again. For example,

- City, State are the same thing and also as we have a unique id's for the school, so we do not need the name of the school in our data.

Also, there are many more variables that I think not having useful information and they are listed below.

```
1 school=school.drop('NAME_OF_SCHOOL','Street Address','City','State','F
2         'CPS Performance Policy Level','Safety Icon','Family Invol
3         'Instruction Icon','Leaders Icon','Teachers Icon','Parent
```

```
1 school.show()
```

School ID	Elementary, Middle, or High School	ZIP Code	Link	Netv
610038	ES	60614	<a href="http://schoolrepo...">http://schoolrepo...</a>	Fullerton
610281	ES	60649	<a href="http://schoolrepo...">http://schoolrepo...</a>	Skyway E
610185	ES	60652	<a href="http://schoolrepo...">http://schoolrepo...</a>	Midway E
609993	ES	60609	<a href="http://schoolrepo...">http://schoolrepo...</a>	Pershing
610513	HS	60609	<a href="http://schoolrepo...">http://schoolrepo...</a>	Southwest
610212	MS	60625	<a href="http://schoolrepo...">http://schoolrepo...</a>	O'Hare E
609720	HS	60618	<a href="http://schoolrepo...">http://schoolrepo...</a>	North-No
610342	ES	60622	<a href="http://schoolrepo...">http://schoolrepo...</a>	Fulton E
610524	HS	60618	<a href="http://schoolrepo...">http://schoolrepo...</a>	North-No
610209	ES	60625	<a href="http://schoolrepo...">http://schoolrepo...</a>	O'Hare E
609799	ES	60618	<a href="http://schoolrepo...">http://schoolrepo...</a>	Ravenswo
609947	ES	60609	<a href="http://schoolrepo...">http://schoolrepo...</a>	Pershing
609963	ES	60657	<a href="http://schoolrepo...">http://schoolrepo...</a>	Ravenswo
610210	ES	60622	<a href="http://schoolrepo...">http://schoolrepo...</a>	Fulton E
609808	ES	60628	<a href="http://schoolrepo...">http://schoolrepo...</a>	Rock Isl
610028	ES	60628	<a href="http://schoolrepo...">http://schoolrepo...</a>	Rock Isl
610098	ES	60651	<a href="http://schoolrepo...">http://schoolrepo...</a>	Garfield
609788	ES	60643	<a href="http://schoolrepo...">http://schoolrepo...</a>	Rock Isl
610334	HS	60624	<a href="http://schoolrepo...">http://schoolrepo...</a>	West Sid
610131	ES	60608	<a href="http://schoolrepo...">http://schoolrepo...</a>	Austin-N

only showing top 20 rows

```
1 school=school.drop('Link ','Network Manager','CPS Performance Policy S
2         'Leaders Icon ','Parent Engagement Icon ')
```

```
1 school.show()
```

School ID	Elementary, Middle, or High School	ZIP Code	Adequate Yearly Progress Made
610038	ES	60614	M
610281	ES	60649	M
610185	ES	60652	M
609993	ES	60609	M
610513	HS	60609	NC
610212	MS	60625	Ye

609720	HS	60618	Yes
610342	ES	60622	No
610524	HS	60618	No
610209	ES	60625	No
609799	ES	60618	No
609947	ES	60609	No
609963	ES	60657	No
610210	ES	60622	No
609808	ES	60628	No
610028	ES	60628	No
610098	ES	60651	No
609788	ES	60643	No
610334	HS	60624	No
610131	ES	60608	No

only showing top 20 rows

```

1  c=0
2  for i,j in school.dtypes:
3      if j=='string':
4          c+=1
5          print(i)
6
7  print()
8  print('----->>>')
9  print('The categorical variables reduced to ----->', c)

```

Elementary, Middle, or High School  
 Adequate Yearly Progress Made?  
 Track Schedule  
 HEALTHY\_SCHOOL\_CERTIFIED  
 Family Involvement Score  
 Leaders Score  
 Teachers Icon  
 Teachers Score  
 Parent Engagement Score  
 Parent Environment Score  
 AVERAGE\_STUDENT\_ATTENDANCE  
 Average Teacher Attendance  
 Individualized Education Program Compliance Rate  
 Pk-2 Literacy %  
 Pk-2 Math %  
 Gr3-5 Grade Level Math %  
 Gr3-5 Grade Level Read %  
 Gr3-5 Keep Pace Read %  
 Gr3-5 Keep Pace Math %  
 Gr6-8 Grade Level Math %  
 Gr6-8 Grade Level Read %  
 Gr6-8 Keep Pace Math %  
 Gr6-8 Keep Pace Read %  
 Gr-8 Explore Math %  
 Gr-8 Explore Read %  
 ISAT Value Add Color Math  
 ISAT Value Add Color Read  
 Students Taking Algebra %



Students Passing Algebra %  
9th Grade EXPLORE (2009)  
9th Grade EXPLORE (2010)  
10th Grade PLAN (2009)  
10th Grade PLAN (2010)  
Net Change EXPLORE and PLAN  
11th Grade Average ACT (2011)  
Net Change PLAN and ACT  
College Eligibility %  
Graduation Rate %  
College Enrollment Rate %  
Freshman on Track Rate %

```
----->>>
The categorical variables reduced to -----> 40
```

Choosing the "Adequate Yearly Progress Made?" variable as a target variables could lead us to answer this research question that *which schoo did have very well progress among the all schools in Chicao?*

```
1  ## chosing =====> Adequate Yearly Progress Made? yes,no as a
```

```
1 school.select('Adequate Yearly Progress Made? ').show()
```

[illegible]

```

1 school=school.withColumnRenamed('Adequate Yearly Progress Made?', 'Target')

1 # from now the Adequate Yearly Progress Made? variable name has changed
2
3 school.select('Target').show()

```

```

+-----+
|Target|
+-----+
|      |
|      |
|      |
|      |
|      |
|NDA   |
|Yes   |
|Yes   |
|No    |
|NDA   |
|No    |
|No    |
|No    |
|No    |
|No    |
|No    |
|No    |
|No    |
|No    |
|No    |
|No    |
+-----+

```

only showing top 20 rows

```

1 school.select('Target').toPandas().value_counts()

```

```

Target
No      476
Yes      72
NDA     18
dtype: int64

```

```

1 school.select('Target').toPandas().describe()

```

	Target
<b>count</b>	566
<b>unique</b>	3
<b>top</b>	No
<b>freq</b>	476

As we see the most frequent variable is our data has responded

```
1 school.select('Target').toPandas().value_counts().to_dict()
```

```
{('NDA',): 18, ('No',): 476, ('Yes',): 72}
```

```
1 school.show(10)
```

```
2
```

```
+-----+-----+-----+-----+-----+
|School ID|Elementary, Middle, or High School|ZIP Code|Target|Track Schedule|HEALTHY_
+-----+-----+-----+-----+-----+
| 610038|ES| 60614|No|Standard|
| 610281|ES| 60649|No|Track_E|
| 610185|ES| 60652|No|Standard|
| 609993|ES| 60609|No|Track_E|
| 610513|HS| 60609|NDA|Standard|
| 610212|MS| 60625|Yes|Standard|
| 609720|HS| 60618|Yes|Standard|
| 610342|ES| 60622|No|Standard|
| 610524|HS| 60618|NDA|Standard|
| 610209|ES| 60625|No|Track_E|
+-----+-----+-----+-----+-----+
```

only showing top 10 rows

```
1 pip install pyjanitor
```

```
Requirement already satisfied: more-itertools<7.0.0,>=4.0.0 in /usr/local/lib/python3.6/dist-packages (4.0.0)
Requirement already satisfied: atomicwrites>=1.0 in /usr/local/lib/python3.6/dist-packages (1.0)
Requirement already satisfied: virtualenv>=20.0.8 in /usr/local/lib/python3.6/dist-packages (20.0.8)
Requirement already satisfied: importlib-metadata; python_version < "3.8" in /usr/local/lib/python3.6/dist-packages (1.6.0)
Requirement already satisfied: nodeenv>=0.11.1 in /usr/local/lib/python3.6/dist-packages (0.11.1)
Requirement already satisfied: importlib-resources; python_version < "3.7" in /usr/local/lib/python3.6/dist-packages (3.2.0)
Requirement already satisfied: identify>=1.0.0 in /usr/local/lib/python3.6/dist-packages (1.0.0)
Requirement already satisfied: cfgv>=2.0.0 in /usr/local/lib/python3.6/dist-packages (2.0.0)
Requirement already satisfied: pyyaml>=5.1 in /usr/local/lib/python3.6/dist-packages (5.1.2)
Requirement already satisfied: nbconvert!=5.4 in /usr/local/lib/python3.6/dist-packages (5.6.1)
Requirement already satisfied: nbformat in /usr/local/lib/python3.6/dist-packages (4.0.1)
Requirement already satisfied: docutils in /usr/local/lib/python3.6/dist-packages (0.14)
Requirement already satisfied: Jinja2 in /usr/local/lib/python3.6/dist-packages (2.10.1)
Requirement already satisfied: sphinx>=1.8 in /usr/local/lib/python3.6/dist-packages (3.4.3)
Requirement already satisfied: ipython>=4.0.0 in /usr/local/lib/python3.6/dist-packages (7.16.0)
Requirement already satisfied: coverage>=4.4 in /usr/local/lib/python3.6/dist-packages (4.5.4)
Requirement already satisfied: sortedcontainers<3.0.0,>=2.1.0 in /usr/local/lib/python3.6/dist-packages (2.1.0)
Requirement already satisfied: decorator in /usr/local/lib/python3.6/dist-packages (4.4.2)
Requirement already satisfied: ipython-genutils in /usr/local/lib/python3.6/dist-packages (0.2.0)
Requirement already satisfied: pyflakes<2.3.0,>=2.2.0 in /usr/local/lib/python3.6/dist-packages (2.2.0)
Requirement already satisfied: mccabe<0.7.0,>=0.6.0 in /usr/local/lib/python3.6/dist-packages (0.6.1)
Requirement already satisfied: pycodestyle<2.7.0,>=2.6.0a1 in /usr/local/lib/python3.6/dist-packages (2.6.0)
Requirement already satisfied: mpmath>=0.19 in /usr/local/lib/python3.6/dist-packages (0.19.0)
Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in /usr/local/lib/python3.6/dist-packages (2.4.7)
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.6/dist-packages (1.0.1)
Requirement already satisfied: cyclo in /usr/local/lib/python3.6/dist-packages (0.10)
Requirement already satisfied: pytz>=2017.2 in /usr/local/lib/python3.6/dist-packages (2017.2)
Requirement already satisfied: filelock<4,>=3.0.0 in /usr/local/lib/python3.6/dist-packages (3.0.0)
Requirement already satisfied: distlib<1,>=0.3.1 in /usr/local/lib/python3.6/dist-packages (0.3.1)
Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.6/dist-packages (0.5.2)
```

```

Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.6/dist-packages
Requirement already satisfied: testpath in /usr/local/lib/python3.6/dist-packages
Requirement already satisfied: bleach in /usr/local/lib/python3.6/dist-packages (
Requirement already satisfied: pandocfilters>=1.4.1 in /usr/local/lib/python3.6/d
Requirement already satisfied: defusedxml in /usr/local/lib/python3.6/dist-packag
Requirement already satisfied: mistune<2,>=0.8.1 in /usr/local/lib/python3.6/dist
Requirement already satisfied: pygments in /usr/local/lib/python3.6/dist-packages
Requirement already satisfied: entrypoints>=0.2.2 in /usr/local/lib/python3.6/dis
Requirement already satisfied: jsonschema!=2.5.0,>=2.4 in /usr/local/lib/python3.
Requirement already satisfied: MarkupSafe>=0.23 in /usr/local/lib/python3.6/dist-
Requirement already satisfied: imagesize in /usr/local/lib/python3.6/dist-package
Requirement already satisfied: requests>=2.0.0 in /usr/local/lib/python3.6/dist-p
Requirement already satisfied: alabaster<0.8,>=0.7 in /usr/local/lib/python3.6/di
Requirement already satisfied: babel!=2.0,>=1.3 in /usr/local/lib/python3.6/dist-
Requirement already satisfied: sphinxcontrib-websupport in /usr/local/lib/python3
Requirement already satisfied: packaging in /usr/local/lib/python3.6/dist-package
Requirement already satisfied: snowballstemmer>=1.1 in /usr/local/lib/python3.6/d
Requirement already satisfied: pexpect; sys_platform != "win32" in /usr/local/lib
Requirement already satisfied: prompt-toolkit<2.0.0,>=1.0.4 in /usr/local/lib/pyt
Requirement already satisfied: simplegeneric>0.8 in /usr/local/lib/python3.6/dist
Requirement already satisfied: pickleshare in /usr/local/lib/python3.6/dist-packa
Requirement already satisfied: webencodings in /usr/local/lib/python3.6/dist-pack
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.6/dis
Requirement already satisfied: urllib3!=1.25.0,!1.25.1,<1.26,>=1.21.1 in /usr/lo
Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.6/dist
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.6/dist-pack
Requirement already satisfied: sphinxcontrib-serializinghtml in /usr/local/lib/py
Requirement already satisfied: ptyprocess>=0.5 in /usr/local/lib/python3.6/dist-p
Requirement already satisfied: wcwidth in /usr/local/lib/python3.6/dist-packages

```

```

1 from pyspark.sql import DataFrame
2 import janitor.spark

```

```

/usr/local/lib/python3.6/dist-packages/distributed/config.py:20: YAMLLoadWarning: cal
defaults = yaml.load(f)

```

Cleaning the data sets are the most important task. In these data sets all the variables names and contents are not properly provided. There are many issues that can affect the queries, so let do cleaning task on both data sets.

```

1 school=school.clean_names()

```

```

1 school.show(5)

```

```

+-----+-----+-----+-----+-----+-----+
|school_id|elementary_middle_or_high_school|zip_code|target|track_schedule|healthy_sc
+-----+-----+-----+-----+-----+-----+

```

610038	ES	60614	No	Standard
610281	ES	60649	No	Track_E
610185	ES	60652	No	Standard
609993	ES	60609	No	Track_E
610513	HS	60609	NDA	Standard

only showing top 5 rows

```
1 school.select('elementary_middle_or_high_school').toPandas().value_counts()
```

```
elementary_middle_or_high_school
ES                                462
HS                                93
MS                                11
dtype: int64
```

```
1 from pyspark.ml.feature import StringIndexer, VectorAssembler, OneHotEncoder
2 from pyspark.sql.functions import when, col
3
```

```
1 school.toPandas().shape
```

```
(566, 59)
```

```
1 census.printSchema()
```

```
root
|-- COMMUNITY_AREA_NUMBER: integer (nullable = true)
|-- PERCENT OF HOUSING CROWDED: double (nullable = true)
|-- PERCENT HOUSEHOLDS BELOW POVERTY: double (nullable = true)
|-- PERCENT AGED 16+ UNEMPLOYED: double (nullable = true)
|-- PERCENT AGED 25+ WITHOUT HIGH SCHOOL DIPLOMA: double (nullable = true)
|-- PERCENT AGED UNDER 18 OR OVER 64: double (nullable = true)
|-- PER_CAPITA_INCOME : integer (nullable = true)
|-- HARDSHIP_INDEX: integer (nullable = true)
```

```
1 census=census.clean_names()
```

```
1 census.show(10)
```

community_area_number	percent_of_housing_crowded	percent_households_below_poverty
1	7.7	23.6
2	7.8	17.2
3	3.8	24.0
4	3.4	10.9
5	0.3	7.5
6	1.1	11.4
7	0.8	12.3

8	1.9	12.9
9	1.1	3.3
10	2.0	5.4

only showing top 10 rows

```
1 # Saving schools as s temp and census as c temp for more sql queries
2 census.createOrReplaceTempView('c')
3 school.createOrReplaceTempView('s')
```

```
1 school.show(5)
```

school_id	elementary_middle_or_high_school	zip_code	target	track_schedule	healthy_sc
610038	ES	60614	No	Standard	
610281	ES	60649	No	Track_E	
610185	ES	60652	No	Standard	
609993	ES	60609	No	Track_E	
610513	HS	60609	NDA	Standard	

only showing top 5 rows

```
1 # Finding which columns are common between two data sets.
2 set(school.columns).intersection(set(census.columns))

{'community_area_number'}
```

```
1 census.toPandas().shape

(77, 8)
```

```
1 # Joining two data sets based on common column
2 data=school.join(census,on='community_area_number',how='inner').distir
```

```
1 data.printSchema()
```

```
root
|-- community_area_number: integer (nullable = true)
|-- school_id: integer (nullable = true)
|-- elementary_middle_or_high_school: string (nullable = true)
|-- zip_code: integer (nullable = true)
|-- target: string (nullable = true)
|-- track_schedule: string (nullable = true)
|-- healthy_school_certified: string (nullable = true)
|-- safety_score: integer (nullable = true)
|-- family_involvement_score: string (nullable = true)
|-- environment_score: integer (nullable = true)
```

```
-- instruction_score: integer (nullable = true)
-- leaders_score_: string (nullable = true)
-- teachers_icon_: string (nullable = true)
-- teachers_score: string (nullable = true)
-- parent_engagement_score: string (nullable = true)
-- parent_environment_score: string (nullable = true)
-- average_student_attendance: string (nullable = true)
-- rate_of_misconducts_per_100_students_: double (nullable = true)
-- average_teacher_attendance: string (nullable = true)
-- individualized_education_program_compliance_rate_: string (nullable = true)
-- pk_2_literacy_%: string (nullable = true)
-- pk_2_math_%: string (nullable = true)
-- gr3_5_grade_level_math_%: string (nullable = true)
-- gr3_5_grade_level_read_%: string (nullable = true)
-- gr3_5_keep_pace_read_%: string (nullable = true)
-- gr3_5_keep_pace_math_%: string (nullable = true)
-- gr6_8_grade_level_math_%: string (nullable = true)
-- gr6_8_grade_level_read_%: string (nullable = true)
-- gr6_8_keep_pace_math%: string (nullable = true)
-- gr6_8_keep_pace_read_%: string (nullable = true)
-- gr_8_explore_math_%: string (nullable = true)
-- gr_8_explore_read_%: string (nullable = true)
-- isat_exceeding_math_%: double (nullable = true)
-- isat_exceeding_reading_%: double (nullable = true)
-- isat_value_add_math: double (nullable = true)
-- isat_value_add_read: double (nullable = true)
-- isat_value_add_color_math: string (nullable = true)
-- isat_value_add_color_read: string (nullable = true)
-- students_taking_algebra_%: string (nullable = true)
-- students_passing_algebra_%: string (nullable = true)
-- 9th_grade_explore_2009_: string (nullable = true)
-- 9th_grade_explore_2010_: string (nullable = true)
-- 10th_grade_plan_2009_: string (nullable = true)
-- 10th_grade_plan_2010_: string (nullable = true)
-- net_change_explore_and_plan: string (nullable = true)
-- 11th_grade_average_act_2011_: string (nullable = true)
-- net_change_plan_and_act: string (nullable = true)
-- college_eligibility_%: string (nullable = true)
-- graduation_rate_%: string (nullable = true)
-- college_enrollment_rate_%: string (nullable = true)
-- college_enrollment: integer (nullable = true)
-- general_services_route_: integer (nullable = true)
-- freshman_on_track_rate_%: string (nullable = true)
-- x_coordinate: double (nullable = true)
-- y_coordinate: double (nullable = true)
-- latitude: double (nullable = true)
-- longitude: double (nullable = true)
-- ward: integer (nullable = true)
```

1 data.limit(5).show()

```
+-----+-----+-----+-----+-----+
|community_area_number|school_id|elementary_middle_or_high_school|zip_code|target|tra
+-----+-----+-----+-----+-----+
|          4|    609852|          ES|    60625|    Yes|
|         10|    609937|          ES|    60656|    No|
|         24|    609828|          ES|    60622|    Yes|
|          6|    610355|          ES|    60613|    Yes|
|         61|    609929|          ES|    60609|    No|
```

```
1 # dropping the na values from the data
2 data=data.na.drop()
```

```
1 data.toPandas().shape

(436, 66)
```

```
1 # Since this col, does not give much information, dropping it.
2 data=data.drop('teachers_icon_')
```

```
1 data.show(5)
```

```
+-----+-----+-----+-----+-----+-----+
|community_area_number|school_id|elementary_middle_or_high_school|zip_code|target|tra
+-----+-----+-----+-----+-----+-----+
|          4|    609852|                                ES|   60625|   Yes|
|         10|    609937|                                ES|   60656|   No|
|         24|    609828|                                ES|   60622|   Yes|
|          6|    610355|                                ES|   60613|   Yes|
|         61|    609929|                                ES|   60609|   No|
+-----+-----+-----+-----+-----+-----+
only showing top 5 rows
```

```
1 data.describe().show()
```

```
+-----+-----+-----+-----+-----+-----+
|summary|community_area_number|          school_id|elementary_middle_or_high_school|
+-----+-----+-----+-----+-----+-----+
|  count|          436|          436|          436|
|   mean|  38.41743119266055|  610068.1880733945| null|66
| stddev|  21.84027848770573|  180.80035714785598| null|21
|   min|          1|    609725|    ES|
|   max|         77|    610544|    MS|
+-----+-----+-----+-----+-----+-----+
```

Counting the rows with the respect of their columns ti find out how

- ▼ many NDA's are in our table. NOTE: The NDA gives us NO information. Let find out!

```
1 from pyspark.sql.functions import count
```



```

1  for i in data.columns:
2      print(data.filter(data[i]=='NDA').count(),'=====>' ,data[i],"=
3  #data.filter(data['students_taking_algebra_%']== 'NDA')

0 =====> Column<b'healthy_school_certified'> =====> Have NDA's
0 =====> Column<b'safety_score'> =====> Have NDA's
204 =====> Column<b'family_involvement_score'> =====> Have NDA's
0 =====> Column<b'environment_score'> =====> Have NDA's
0 =====> Column<b'instruction_score'> =====> Have NDA's
206 =====> Column<b'leaders_score_'> =====> Have NDA's
206 =====> Column<b'teachers_score'> =====> Have NDA's
78 =====> Column<b'parent_engagement_score'> =====> Have NDA's
78 =====> Column<b'parent_environment_score'> =====> Have NDA's
0 =====> Column<b'average_student_attendance'> =====> Have NDA's
0 =====> Column<b'rate_of_misconducts_per_100_students_'> =====> Hav
0 =====> Column<b'average_teacher_attendance'> =====> Have NDA's
0 =====> Column<b'individualized_education_program_compliance_rate_'> ====
68 =====> Column<b'pk_2_literacy_%'> =====> Have NDA's
130 =====> Column<b'pk_2_math_%'> =====> Have NDA's
22 =====> Column<b'gr3_5_grade_level_math_%'> =====> Have NDA's
22 =====> Column<b'gr3_5_grade_level_read_%'> =====> Have NDA's
22 =====> Column<b'gr3_5_keep_pace_read_%'> =====> Have NDA's
22 =====> Column<b'gr3_5_keep_pace_math_%'> =====> Have NDA's
9 =====> Column<b'gr6_8_grade_level_math_%'> =====> Have NDA's
8 =====> Column<b'gr6_8_grade_level_read_%'> =====> Have NDA's
9 =====> Column<b'gr6_8_keep_pace_math%'> =====> Have NDA's
8 =====> Column<b'gr6_8_keep_pace_read_%'> =====> Have NDA's
27 =====> Column<b'gr_8_explore_math_%'> =====> Have NDA's
27 =====> Column<b'gr_8_explore_read_%'> =====> Have NDA's
0 =====> Column<b'isat_exceeding_math_%'> =====> Have NDA's
0 =====> Column<b'isat_exceeding_reading_%'> =====> Have NDA's
0 =====> Column<b'isat_value_add_math'> =====> Have NDA's
0 =====> Column<b'isat_value_add_read'> =====> Have NDA's
0 =====> Column<b'isat_value_add_color_math'> =====> Have NDA's
0 =====> Column<b'isat_value_add_color_read'> =====> Have NDA's
284 =====> Column<b'students_taking_algebra_%'> =====> Have NDA's
312 =====> Column<b'students_passing_algebra_%'> =====> Have NDA's
427 =====> Column<b'9th_grade_explore_2009_'> =====> Have NDA's
428 =====> Column<b'9th_grade_explore_2010_'> =====> Have NDA's
429 =====> Column<b'10th_grade_plan_2009_'> =====> Have NDA's
428 =====> Column<b'10th_grade_plan_2010_'> =====> Have NDA's
428 =====> Column<b'net_change_explore_and_plan'> =====> Have NDA's
429 =====> Column<b'11th_grade_average_act_2011_'> =====> Have NDA's
429 =====> Column<b'net_change_plan_and_act'> =====> Have NDA's
429 =====> Column<b'college_eligibility_%'> =====> Have NDA's
430 =====> Column<b'graduation_rate_%'> =====> Have NDA's
430 =====> Column<b'college_enrollment_rate_%'> =====> Have NDA's
0 =====> Column<b'college_enrollment'> =====> Have NDA's
0 =====> Column<b'general_services_route_'> =====> Have NDA's
428 =====> Column<b'freshman_on_track_rate_%'> =====> Have NDA's
0 =====> Column<b'x_coordinate'> =====> Have NDA's
0 =====> Column<b'y_coordinate'> =====> Have NDA's

0 =====> Column<b'latitude'> =====> Have NDA's
0 =====> Column<b'longitude'> =====> Have NDA's
0 =====> Column<b'ward'> =====> Have NDA's
0 =====> Column<b'police_district'> =====> Have NDA's
0 =====> Column<b'percent_of_housing_crowded'> =====> Have NDA's
0 =====> Column<b'percent_households_below_poverty'> =====> Have NDA
0 =====> Column<b'percent aged 16+ unemployed'> =====> Have NDA's

```

```

0 =====>> Column<b'percent_aged_25+_without_high_school_diploma'> =====
0 =====>> Column<b'percent_aged_under_18_or_over_64'> =====> Have NDA
0 =====>> Column<b'per_capita_income_'> =====> Have NDA's

```

- ▼ Since some columns have many NDA's it is better to drop them

Since does not give any useful information

```

1 # these cols have many NDA as results of the above calculation and sir
2 to_be_dropped=['family_involvement_score','leaders_score_',
3               'teachers_score','parent_engagement_score',
4               'parent_environment_score','pk_2_literacy_%',
5               'pk_2_math_%','gr3_5_grade_level_math_%',
6               'gr3_5_grade_level_read_%','gr3_5_keep_pace_read_%',
7               'gr3_5_keep_pace_math_%','gr6_8_grade_level_math_%',
8               'gr6_8_grade_level_read_%','gr6_8_keep_pace_math%',
9               'gr6_8_keep_pace_read_%','gr_8_explore_math_%',
10              'gr_8_explore_read_%','students_taking_algebra_%',
11              'students_passing_algebra_%','9th_grade_explore_2009_',
12              '9th_grade_explore_2010_','10th_grade_plan_2009_',
13              '10th_grade_plan_2010_','net_change_explore_and_plan',
14              '11th_grade_average_act_2011_','net_change_plan_and_act',
15              'college_eligibility_%','graduation_rate_%',
16              'college_enrollment_rate_%','freshman_on_track_rate_%',
17              ]

```

```
1 data_NDA_drop=data.drop(*to_be_dropped)
```

```
1 # Data reduced to 35 columns with the 436 records
```

```
2 data_NDA_drop.toPandas().shape
```

```
(436, 35)
```

```
1 len(data_NDA_drop.dtypes)
```

```
35
```

- ▼ Checking out that how many categorical variables are in our data

```

1 ## These columns has the string type and they should be converted to
2 for i,j in data_NDA_drop.dtypes:
3     if j=='string':
4         print(i '=====> has a string type of a value should be nume

```

```

4 print(1, -----, has a string type of a value should be numeric
5 print()
6 print('---'*50)

```

elementary\_middle\_or\_high\_school =====> has a string type of a value should be numeric

-----  
 target =====> has a string type of a value should be numeric type

-----  
 track\_schedule =====> has a string type of a value should be numeric type

-----  
 healthy\_school\_certified =====> has a string type of a value should be numeric

-----  
 average\_student\_attendance =====> has a string type of a value should be numeric

-----  
 average\_teacher\_attendance =====> has a string type of a value should be numeric

-----  
 individualized\_education\_program\_compliance\_rate\_ =====> has a string type of

-----  
 isat\_value\_add\_color\_math =====> has a string type of a value should be numeric

-----  
 isat\_value\_add\_color\_read =====> has a string type of a value should be numeric

1 data\_NDA\_drop.select('elementary\_middle\_or\_high\_school').show()

```

+-----+
|elementary_middle_or_high_school|
+-----+
|                                ES|
|                                ES|
|                                ES|
|                                ES|
|                                ES|
|                                ES|
|                                ES|
|                                ES|
|                                ES|
|                                ES|
|                                ES|
|                                HS|
|                                ES|
|                                ES|
|                                ES|
|                                MS|
|                                ES|
|                                ES|
|                                ES|
|                                ES|
+-----+

```

only showing top 20 rows

```
1 data_NDA_drop.select('elementary_middle_or_high_school').toPandas().value_counts()

elementary_middle_or_high_school
ES                                417
MS                                 11
HS                                 8
dtype: int64
```

## Encoding the categorical variable to dummies process doing with StringIndexer function from Mlib

```
1 indexer=StringIndexer(inputCol='elementary_middle_or_high_school',outputCol='elementary_middle_or_high_school_encoded')

1 indexed=indexer.fit(data_NDA_drop).transform(data_NDA_drop)

1 indexed.show(10)
```

community_area_number	school_id	elementary_middle_or_high_school	zip_code	target	track_schedule
4	609852	ES	60625	Yes	Standard
10	609937	ES	60656	No	Standard
24	609828	ES	60622	Yes	Standard
6	610355	ES	60613	Yes	Standard
61	609929	ES	60609	No	Standard
28	610180	ES	60608	No	Standard
61	610239	ES	60609	No	Standard
71	609805	ES	60620	No	Standard
61	610167	ES	60609	No	Standard
69	609813	ES	60637	No	Standard

only showing top 10 rows

```
1 indexed=indexed.drop('elementary_middle_or_high_school')

1 indexed=indexed.withColumnRenamed('elementary_middle_or_high_school_encoded','elementary_middle_or_high_school_encoded')

1 indexed.show(10)
```

community_area_number	school_id	zip_code	target	track_schedule	healthy_school_certificate
4	609852	60625	Yes	Standard	
10	609937	60656	No	Standard	

24	609828	60622	Yes	Track_E
6	610355	60613	Yes	Standard
61	609929	60609	No	Track_E
28	610180	60608	No	Track_E
61	610239	60609	No	Track_E
71	609805	60620	No	Track_E
61	610167	60609	No	Standard
69	609813	60637	No	Track_E

```
+-----+-----+-----+-----+-----+
only showing top 10 rows
```

```
1 indexed.select('track_schedule').toPandas().value_counts()
```

```
track_schedule
Standard      237
Track_E       199
dtype: int64
```

```
1 indexer=StringIndexer(inputCol='track_schedule',outputCol='track_sche
```

```
1 indexed=indexer.fit(indexed).transform(indexed)
```

```
1 indexed.show(5)
```

community_area_number	school_id	zip_code	target	track_schedule	healthy_school_certifi
4	609852	60625	Yes	Standard	
10	609937	60656	No	Standard	
24	609828	60622	Yes	Track_E	
6	610355	60613	Yes	Standard	
61	609929	60609	No	Track_E	

```
+-----+-----+-----+-----+-----+
only showing top 5 rows
```

```
1 indexed=indexed.withColumn('track_schedule',indexed['track_schedule_ir
```

```
1 indexed.show(4)
```

```
2 indexed=indexed.drop('track_schedule_indexer')
```

```
3 indexed.show(4)
```

community_area_number	school_id	zip_code	target	track_schedule	healthy_school_certifi
4	609852	60625	Yes	0.0	
10	609937	60656	No	0.0	
24	609828	60622	Yes	1.0	
6	610355	60613	Yes	0.0	

```

+-----+-----+-----+-----+-----+
only showing top 4 rows

+-----+-----+-----+-----+-----+
|community_area_number|school_id|zip_code|target|track_schedule|healthy_school_certifi
+-----+-----+-----+-----+-----+
|          4|    609852|   60625|   Yes|          0.0|
|         10|    609937|   60656|    No|          0.0|
|         24|    609828|   60622|   Yes|          1.0|
|          6|    610355|   60613|   Yes|          0.0|
+-----+-----+-----+-----+-----+
only showing top 4 rows

```

```
1 indexed.select('healthy_school_certified').toPandas().value_counts()
```

```

healthy_school_certified
No                        424
Yes                       12
dtype: int64

```

```
1 indexer=StringIndexer(inputCol='healthy_school_certified',outputCol='l
```

```
1 indexed=indexer.fit(indexed).transform(indexed)
```

```
1 indexed.show(50)
```

```

+-----+-----+-----+-----+-----+
|community_area_number|school_id|zip_code|target|track_schedule|healthy_school_certifi
+-----+-----+-----+-----+-----+
|          4|    609852|   60625|   Yes|          0.0|
|         10|    609937|   60656|    No|          0.0|
|         24|    609828|   60622|   Yes|          1.0|
|          6|    610355|   60613|   Yes|          0.0|
|         61|    609929|   60609|    No|          1.0|
|         28|    610180|   60608|    No|          1.0|
|         61|    610239|   60609|    No|          1.0|
|         71|    609805|   60620|    No|          1.0|
|         61|    610167|   60609|    No|          0.0|
|         69|    609813|   60637|    No|          1.0|
|         30|    609973|   60623|    No|          1.0|
|         39|    609746|   60615|    No|          0.0|
|         58|    610353|   60632|    No|          1.0|
|         55|    609856|   60633|    No|          0.0|
|         24|    610076|   60647|    No|          1.0|
|         63|    610532|   60632|    No|          1.0|
|         27|    610251|   60612|    No|          1.0|
|          5|    610010|   60657|    No|          0.0|
|         31|    610125|   60608|    No|          0.0|
|         43|    610103|   60649|    No|          0.0|
|         25|    610367|   60644|    No|          1.0|
|         52|    610198|   60617|    No|          0.0|
|         40|    609819|   60615|    No|          1.0|
|         28|    609812|   60612|    No|          0.0|
|         25|    610092|   60644|    No|          1.0|

```

73	610362	60628	No	1.0
70	609879	60652	No	0.0
24	610073	60612	Yes	0.0
61	609885	60609	No	0.0
35	610110	60653	No	1.0
22	610138	60647	No	0.0
29	609851	60608	No	1.0
6	609850	60613	No	0.0
53	610224	60628	No	1.0
66	610347	60636	No	1.0
73	610027	60620	No	0.0
43	609815	60649	No	0.0
6	609974	60657	Yes	0.0
29	610131	60608	No	1.0
49	610188	60628	No	1.0
10	609995	60646	No	0.0
66	610396	60629	No	0.0
68	610339	60621	No	1.0
17	609810	60634	No	0.0
66	610057	60629	No	1.0
53	610160	60628	No	1.0
38	609781	60609	No	0.0
13	610127	60625	No	0.0
69	610233	60621	No	1.0
44	610093	60619	No	0.0

only showing top 50 rows

```
1 indexed=indexed.withColumn('healthy_school_certified',indexed['healthy
2 indexed=indexed.drop('healthy_school_certified_indexer')
```

```
1 indexed.show(5)
```

community_area_number	school_id	zip_code	target	track_schedule	healthy_school_certified
4	609852	60625	Yes	0.0	
10	609937	60656	No	0.0	
24	609828	60622	Yes	1.0	
6	610355	60613	Yes	0.0	
61	609929	60609	No	1.0	

only showing top 5 rows

- There are very inconvenience number format with the % sign in front of them also they are string type that they should be numeric.
- Procedure to clean these columns as belowe

```
1 indexed.select('average_student_attendance').show()
```

```
+-----+
|average_student_attendance|
+-----+
|          95.10%|
|          95.30%|
|          95.90%|
|          94.90%|
|          95.10%|
|          92.40%|
|          94.90%|
|          95.00%|
|          95.70%|
|          92.80%|
|          96.60%|
|          88.40%|
|          96.50%|
|          95.20%|
|          93.00%|
|          95.50%|
|          94.90%|
|          94.70%|
|          95.50%|
|          92.10%|
+-----+
only showing top 20 rows
```

```
1 from pyspark.sql.functions import format_number,format_string,split
2
```

## ▼ Getting rid of the % percentage sign stuck to the number

```
1 indexed=indexed.withColumn('average_student_attendance',split(indexed|
```

```
1 indexed.select('average_student_attendance').show()
```

```
+-----+
|average_student_attendance|
+-----+
|          95.10|
|          95.30|
|          95.90|
|          94.90|
|          95.10|
|          92.40|
|          94.90|
|          95.00|
|          95.70|
|          92.80|
|          96.60|
|          88.40|
|          96.50|
|          95.20|
```



```
|          93.00|
|          95.50|
|          94.90|
|          94.70|
|          95.50|
|          92.10|
```

```
+-----+
```

```
only showing top 20 rows
```

## ▼ Casting the string format to float format

```
1 indexed=indexed.withColumn('average_student_attendance',indexed['aver
```

```
1 indexed.select('average_student_attendance').dtypes
```

```
[('average_student_attendance', 'float')]
```

```
1 indexed.select('average_teacher_attendance').dtypes
```

```
[('average_teacher_attendance', 'string')]
```

```
1 indexed.select('average_teacher_attendance').show(4)
```

```
+-----+
|average_teacher_attendance|
```

```
+-----+
```

```
|          96.70%|
```

```
|          96.50%|
```

```
|          96.90%|
```

```
|          94.20%|
```

```
+-----+
```

```
only showing top 4 rows
```

## ▼ Getting rid of the % sign and transforming that to a numebr

```
1 indexed=indexed.withColumn('average_teacher_attendance',
2                             split(indexed['average_teacher_attendance'],'%').get
```

```
1 indexed.select('average_teacher_attendance').show(10)
```

```
+-----+
|average_teacher_attendance|
```

```
+-----+
```

```
|          96.70|
```

```
|          96.50|
```

```
|          96.90|
```

```
|          94.20|
```

```
|          96.80|
|          93.30|
|          96.00|
|          94.60|
|          95.70|
|          95.50|
```

```
+-----+
```

```
only showing top 10 rows
```

```
1 indexed=indexed.withColumn('average_teacher_attendance',indexed['average_teacher_attendance'])
```

```
1 indexed.select('average_teacher_attendance').dtypes
```

```
[('average_teacher_attendance', 'float')]
```

```
1 indexed.select('individualized_education_program_compliance_rate_').show()
```

```
+-----+
|individualized_education_program_compliance_rate_|
+-----+
|          98.90%|
|          97.30%|
|          100.00%|
|          100.00%|
|          100.00%|
```

```
+-----+
```

```
only showing top 5 rows
```

```
1 indexed=indexed.withColumn('individualized_education_program_compliance_rate_',
```

```
2                               split(indexed['individualized_education_program_compliance_rate_'],
```

```
1 indexed.select('individualized_education_program_compliance_rate_').show()
```

```
+-----+
|individualized_education_program_compliance_rate_|
+-----+
|          98.90|
|          97.30|
|          100.00|
|          100.00|
|          100.00|
```

```
+-----+
```

```
only showing top 5 rows
```

▼ Categorical variables are useless for ML models, so they should be numeric

```
1 indexed.select('isat_value add color math').show()
```

```
1 indexed.select('isat_value_add_color_math').show()
```

isat_value_add_color_math
Yellow
Yellow
Yellow
Red
Green
Green
Green
Yellow
Red
Green
Yellow
Red
Red
Yellow
Green
Red
Green
Yellow
Red
Yellow

only showing top 20 rows

## ▼ Transforming the SAT values to numeric encoding

```
1 indexer=StringIndexer(inputCol='isat_value_add_color_math',outputCol='
2 indexed=indexer.fit(indexed).transform(indexed)
3
```

```
1 indexed.show(5)
```

community_area_number	school_id	zip_code	target	track_schedule	healthy_school_certificate
4	609852	60625	Yes	0.0	
10	609937	60656	No	0.0	
24	609828	60622	Yes	1.0	
6	610355	60613	Yes	0.0	
61	609929	60609	No	1.0	

only showing top 5 rows

```
1 indexed=indexed.drop('isat_value_add_color_math')
```

```
1 indexed=indexed.withColumnRenamed('isat_value_add_color_math_indexer',
```

```
1 indexed.select('individualized_education_program_compliance_rate_').st
```

```
+-----+
|individualized_education_program_compliance_rate_|
+-----+
|                                                    98.90|
|                                                    97.30|
|                                                    100.00|
|                                                    100.00|
|                                                    100.00|
+-----+
```

only showing top 5 rows

▼ This variable is also string and it should be transformd to the float number

```
1 indexed.select('individualized_education_program_compliance_rate_').dt
```

```
[('individualized_education_program_compliance_rate_', 'string')]
```

```
1 indexed=indexed.withColumn('individualized_education_program_compliance_rate_',
```

```
2 indexed['individualized_education_program_compliance_rate_'].cast(FloatType))
```

```
1 indexed.select('individualized_education_program_compliance_rate_').dt
```

```
[('individualized_education_program_compliance_rate_', 'float')]
```

▼ Finding wich variable is remaining as n string

```
1 for i,j in indexed.dtypes:
2     if j=='string':
3         print(i,'=====> string type yet!')
4         print()
```

```
target =====> string type yet!
```

```
isat_value_add_color_read =====> string type yet!
```

```
1 indexed.select('isat_value_add_color_read').show(5)
```

```
+-----+
|isat_value_add_color_read|
+-----+
|                          Green|
```

```
|          Yellow|
|          Yellow|
|          Yellow|
|           Red |
+-----+
only showing top 5 rows
```

## ▼ Transforming, encoding the variable isat\_value\_add\_color\_read

```
1 indexer=StringIndexer(inputCol='isat_value_add_color_read',outputCol='
2 indexed=indexer.fit(indexed).transform(indexed)
```

```
1 indexed.show(5)
```

```
+-----+-----+-----+-----+-----+-----+
|community_area_number|school_id|zip_code|target|track_schedule|healthy_school_certi|
+-----+-----+-----+-----+-----+-----+
|          4|    609852|   60625|   Yes|          0.0|
|         10|    609937|   60656|    No|          0.0|
|         24|    609828|   60622|   Yes|          1.0|
|          6|    610355|   60613|   Yes|          0.0|
|         61|    609929|   60609|    No|          1.0|
+-----+-----+-----+-----+-----+-----+
only showing top 5 rows
```

```
1 indexed=indexed.drop('isat_value_add_color_read')
```

```
1 indexed=indexed.withColumnRenamed('isat_value_add_color_read_indexer',
```

## ▼ All the variables tranfformed to numerical with the encoding them

```
1 indexed.dtypes
```

```
[('community_area_number', 'int'),
 ('school_id', 'int'),
 ('zip_code', 'int'),
 ('target', 'string'),
 ('track_schedule', 'double'),
 ('healthy_school_certified', 'double'),
 ('safety_score', 'int'),
 ('environment_score', 'int'),
 ('instruction_score', 'int'),
 ('average_student_attendance', 'float'),
 ('rate_of_misconducts_per_100_students_', 'double'),
 ('average_teacher_attendance', 'float'),
 ('individualized_education_program_compliance_rate_', 'float'),
```

```
(('isat_exceeding_math_', 'double'),
 ('isat_exceeding_reading_', 'double'),
 ('isat_value_add_math', 'double'),
 ('isat_value_add_read', 'double'),
 ('college_enrollment', 'int'),
 ('general_services_route_', 'int'),
 ('x_coordinate', 'double'),
 ('y_coordinate', 'double'),
 ('latitude', 'double'),
 ('longitude', 'double'),
 ('ward', 'int'),
 ('police_district', 'int'),
 ('percent_of_housing_crowded', 'double'),
 ('percent_households_below_poverty', 'double'),
 ('percent_aged_16+_unemployed', 'double'),
 ('percent_aged_25+_without_high_school_diploma', 'double'),
 ('percent_aged_under_18_or_over_64', 'double'),
 ('per_capita_income_', 'int'),
 ('hardship_index', 'int'),
 ('elementary_middle_or_high_school', 'double'),
 ('isat_value_add_color_math', 'double'),
 ('isat_value_add_color_read', 'double'])
```

```
1 indexed.select('target').toPandas().value_counts()
```

```
target
No      370
Yes      63
NDA       3
dtype: int64
```

There are three record in our target columns that should be fixed. I

- will use the frequency method to replace the NDA's with the most frequent recods shown in the data.

```
1 indexed.filter(indexed['target']=='NDA').show()
```

```
+-----+-----+-----+-----+-----+-----+
|community_area_number|school_id|zip_code|target|track_schedule|healthy_school_certifi|
+-----+-----+-----+-----+-----+-----+
|                28|   610075|   60608|   NDA|             1.0|
|                24|   610085|   60642|   NDA|             0.0|
|                48|   610280|   60617|   NDA|             0.0|
+-----+-----+-----+-----+-----+-----+
```

- Viewing that how the target variabl's records are showing up in our data set.

```
1 indexed.filter((indexed['community_area_number']==28) |(indexed['commu
```

community_area_number	school_id	zip_code	target	track_schedule	healthy_school_certifi
24	609828	60622	Yes	1.0	
28	610180	60608	No	1.0	
24	610076	60647	No	1.0	
28	609812	60612	No	0.0	
24	610073	60612	Yes	0.0	
28	610075	60608	NDA	1.0	
24	610107	60622	No	0.0	
28	610023	60612	No	1.0	
28	610121	60612	No	1.0	
28	610009	60607	Yes	0.0	
24	610031	60622	No	1.0	
28	609989	60612	No	1.0	
48	610218	60617	No	0.0	
24	610320	60622	No	0.0	
24	610085	60642	NDA	0.0	
28	610177	60607	Yes	0.0	
24	610342	60622	No	0.0	
24	610210	60622	No	0.0	
24	610313	60622	No	0.0	
24	610529	60622	Yes	0.0	

only showing top 20 rows

```
1 # Replacing the NDA's with the None values then changing them. Spark a
2 # This is an easy trick to doing so.
3
4 indexed=indexed.withColumn('target',when(indexed['target']=='NDA',None
```

```
1 indexed.select('target').toPandas().isnull().sum()
```

```
target      3
dtype: int64
```

Because the top most frequency value is No in target variable I will replace the 3 null values with 0

```
1 indexed=indexed.fillna('No')
```

```
1 indexed.select('target').toPandas().isnull().sum()
```

```
target      0
dtype: int64
```

## ▼ Encoding the target variable

```
1 indexer=StringIndexer(inputCol='target',outputCol='target_indexer',)
2 indexed=indexer.fit(indexed).transform(indexed)
```

```
1 indexed.toPandas().isnull().sum()
```

```
community_area_number      0
school_id                  0
zip_code                   0
target                     0
track_schedule             0
healthy_school_certified   0
safety_score              0
environment_score         0
instruction_score          0
average_student_attendance 0
rate_of_misconducts_per_100_students_ 0
average_teacher_attendance 0
individualized_education_program_compliance_rate_ 0
isat_exceeding_math_%      0
isat_exceeding_reading_%   0
isat_value_add_math        0
isat_value_add_read        0
college_enrollment        0
general_services_route_    0
x_coordinate               0
y_coordinate               0
latitude                   0
longitude                  0
ward                      0
police_district            0
percent_of_housing_crowded 0
percent_households_below_poverty 0
percent_aged_16+_unemployed 0
percent_aged_25+_without_high_school_diploma 0
percent_aged_under_18_or_over_64 0
per_capita_income_        0
hardship_index            0
elementary_middle_or_high_school 0
isat_value_add_color_math  0
isat_value_add_color_read  0
target_indexer             0
dtype: int64
```

```
1 indexed.show(10)
```

```
+-----+-----+-----+-----+-----+-----+
|community_area_number|school_id|zip_code|target|track_schedule|healthy_school_certifi|
+-----+-----+-----+-----+-----+-----+
|                4|    609852|    60625|    Yes|            0.0|
|                10|    609937|    60656|     No|            0.0|
|                24|    609828|    60622|     Yes|            1.0|
```



6	610355	60613	Yes	0.0
61	609929	60609	No	1.0
28	610180	60608	No	1.0
61	610239	60609	No	1.0
71	609805	60620	No	1.0
61	610167	60609	No	0.0
69	609813	60637	No	1.0

+-----+  
only showing top 10 rows

## More Queries about finding the interesting facts from our data

Q6. What was the maximum amount of the Per Capita Income among the all schools in Chicago?

```
1 indexed.createOrReplaceTempView('inx')
2 spark.sql("select max(per_capita_income_) from inx").show()
3
```

```
+-----+
|max(per_capita_income_)|
+-----+
|                88669|
+-----+
```

## Q7. Interesting question

Did students with the highest per capita income rate do well performance in their education ? 75 % did NOT and 25 % did.

```
1 spark.sql('select target from inx where per_capita_income_==88669').show()
```

```
+-----+
|target|
+-----+
|    No|
|    No|
|    No|
|   Yes|
+-----+
```

Q8. What was the average percentage of housing population? 5 % per house.

```
1 spark.sql('select avg(percent_of_housing_crowded) from inx').show()
```

```
+-----+
|avg(percent_of_housing_crowded)|
+-----+
|                    5.415825688073393|
+-----+
```

Q9. How many students did enroll among the schools in the state in total? 250629 students.

```
1 spark.sql('select sum(college_enrollment) from inx').show()
```

```
+-----+
|sum(college_enrollment)|
+-----+
|                    250629|
+-----+
```

Q10. How many distinct police districts were found in data? 436

```
1 spark.sql('select distinct count(police_district) from inx').show()
```

```
+-----+
|count(police_district)|
+-----+
|                    436|
+-----+
```

```
1 indexed.show(5)
```

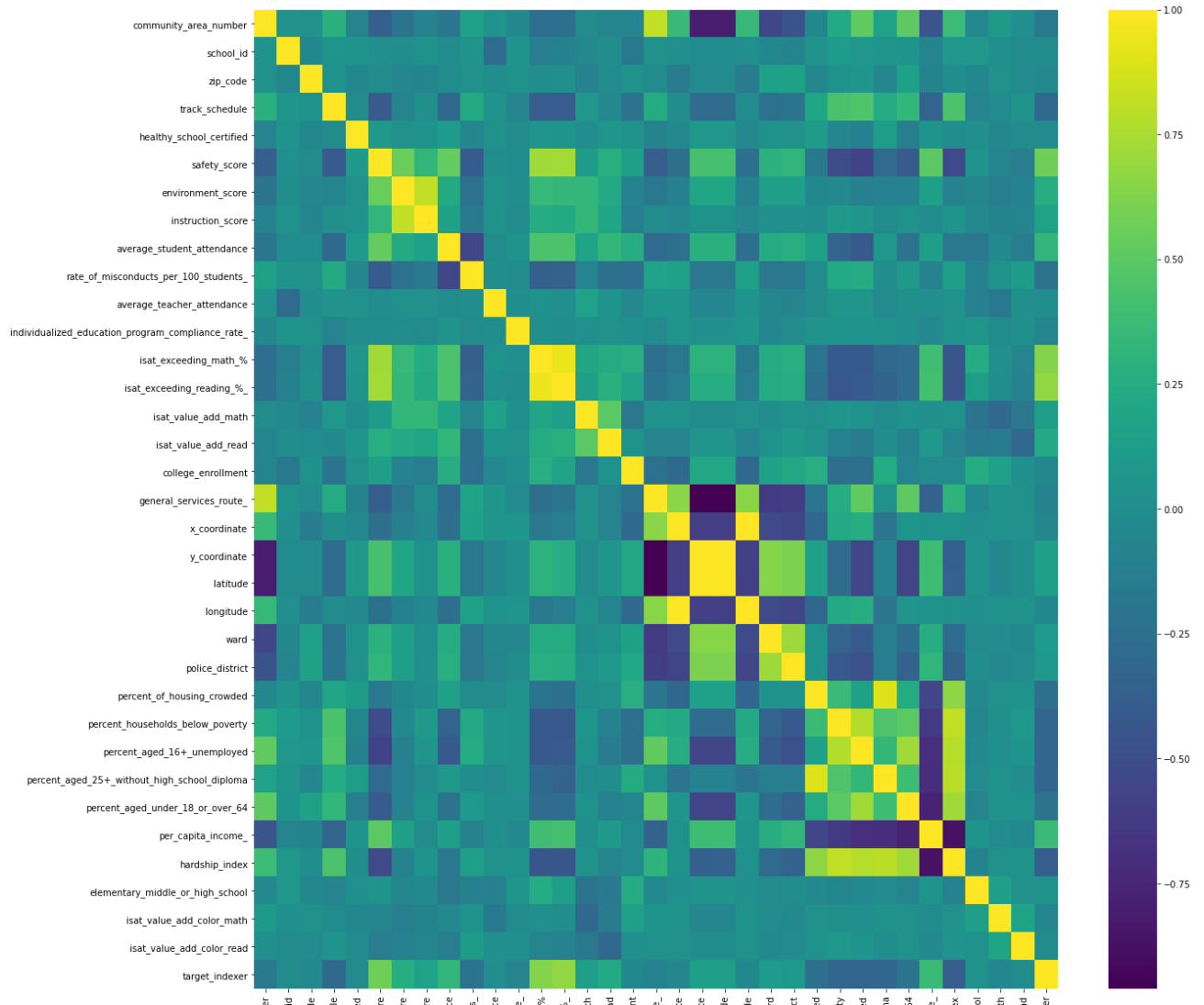
```
+-----+-----+-----+-----+-----+-----+
|community_area_number|school_id|zip_code|target|track_schedule|healthy_school_certificate|
+-----+-----+-----+-----+-----+-----+
|                    4|   609852|   60625|   Yes|              0.0|
|                   10|   609937|   60656|    No|              0.0|
|                   24|   609828|   60622|   Yes|              1.0|
|                    6|   610355|   60613|   Yes|              0.0|
|                   61|   609929|   60609|    No|              1.0|
+-----+-----+-----+-----+-----+-----+
```

only showing top 5 rows

- ▼ Findig the coefficient correlations with visualization to Feature Extraction in the data after joining both data sets.

```
1 plt.figure(figsize=(20,20))
2 sns.heatmap(indexed.toPandas().corr(),cmap='viridis',)
3
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f8bf1d297b8>



As we see visually finding the most efficient features are hard to determin

We need more suffisticated procedures to pick our feartures.

```
1 from pyspark.ml.feature import RFormula
2
```

```
1 formula = RFormula(
2     formula="target ~ .",
3     featuresCol="features",
4     labelCol="label")
```

```
1 output = formula.fit(indexed).transform(indexed)
2 output.select("features", "label").show()
```

```
+-----+-----+
|          features|label|
```

```

+-----+-----+
| [4.0,609852.0,606...| 1.0|
| [10.0,609937.0,60...| 0.0|
| [24.0,609828.0,60...| 1.0|
| [6.0,610355.0,606...| 1.0|
| [61.0,609929.0,60...| 0.0|
| [28.0,610180.0,60...| 0.0|
| [61.0,610239.0,60...| 0.0|
| [71.0,609805.0,60...| 0.0|
| [61.0,610167.0,60...| 0.0|
| [69.0,609813.0,60...| 0.0|
| [30.0,609973.0,60...| 0.0|
| [39.0,609746.0,60...| 0.0|
| [58.0,610353.0,60...| 0.0|
| [55.0,609856.0,60...| 0.0|
| [24.0,610076.0,60...| 0.0|
| [63.0,610532.0,60...| 0.0|
| [27.0,610251.0,60...| 0.0|
| [5.0,610010.0,606...| 0.0|
| [31.0,610125.0,60...| 0.0|
| [43.0,610103.0,60...| 0.0|
+-----+-----+
only showing top 20 rows

```

```
1 output.summary().show()
```

```

+-----+-----+-----+-----+-----+-----+
|summary|community_area_number|school_id|zip_code|target|track|
+-----+-----+-----+-----+-----+-----+
| count|436|436|436|436| |
| mean|38.41743119266055|610068.1880733945|60630.05275229358|null|0.45642201|
| stddev|21.84027848770573|180.80035714785598|21.77258566776887|null|0.49866953|
| min|1|609725|60605|No|
| 25%|23|609918|60618|null|
| 50%|34|610067|60625|null|
| 75%|60|610200|60639|null|
| max|77|610544|60827|Yes|
+-----+-----+-----+-----+-----+-----+

```

Feature Extraction and Selection on the best scores for both data sets, procedure.

Setting the explanatory and response variables to a panda data fram. Later on I'll use the sklearn library to doing so.

```

1 y=indexed.select('target').toPandas()
2
3

```

3 y

	target
0	Yes
1	No
2	Yes
3	Yes
4	No
...	...
431	Yes
432	Yes
433	No
434	Yes
435	No

436 rows × 1 columns

```

1 X=indexed.select('*').drop('target').toPandas()
2 X
3

```

	community_area_number	school_id	zip_code	track_schedule	healthy_school_cert
0	4	609852	60625	0.0	
1	10	609937	60656	0.0	
2	24	609828	60622	1.0	
3	6	610355	60613	0.0	
4	61	609929	60609	1.0	
...	...	...	...	...	
431	48	610316	60617	1.0	
432	2	610191	60659	0.0	
433	68	610173	60621	1.0	
434	24	609863	60622	0.0	
435	15	610179	60634	0.0	

436 rows × 35 columns

```

1 from sklearn.ensemble import ExtraTreesClassifier

```

```
1 model=ExtraTreesClassifier()
```

```
1 model.fit(X,y)
```

```
/usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:1: DataConversionWarning
    """Entry point for launching an IPython kernel.
```

```
ExtraTreesClassifier(bootstrap=False, ccp_alpha=0.0, class_weight=None,
                      criterion='gini', max_depth=None, max_features='auto',
                      max_leaf_nodes=None, max_samples=None,
                      min_impurity_decrease=0.0, min_impurity_split=None,
                      min_samples_leaf=1, min_samples_split=2,
                      min_weight_fraction_leaf=0.0, n_estimators=100,
                      n_jobs=None, oob_score=False, random_state=None, verbose=0,
                      warm_start=False)
```

```
1 model.feature_importances_
```

```
array([0.00454804, 0.00633722, 0.00607385, 0.01697049, 0.00058912,
        0.04865729, 0.00885919, 0.00691484, 0.01871791, 0.00871088,
        0.00539243, 0.00710986, 0.05613585, 0.07849525, 0.00806118,
        0.00823698, 0.0093317 , 0.00526295, 0.00541539, 0.00764897,
        0.00721356, 0.00576654, 0.00565357, 0.00502916, 0.01118319,
        0.01577276, 0.01416577, 0.01511919, 0.00820842, 0.01409268,
        0.02018348, 0.0026574 , 0.00653247, 0.00499776, 0.54595469])
```

```
1 feat_importance=pd.Series(model.feature_importances_,index=X.columns)
```

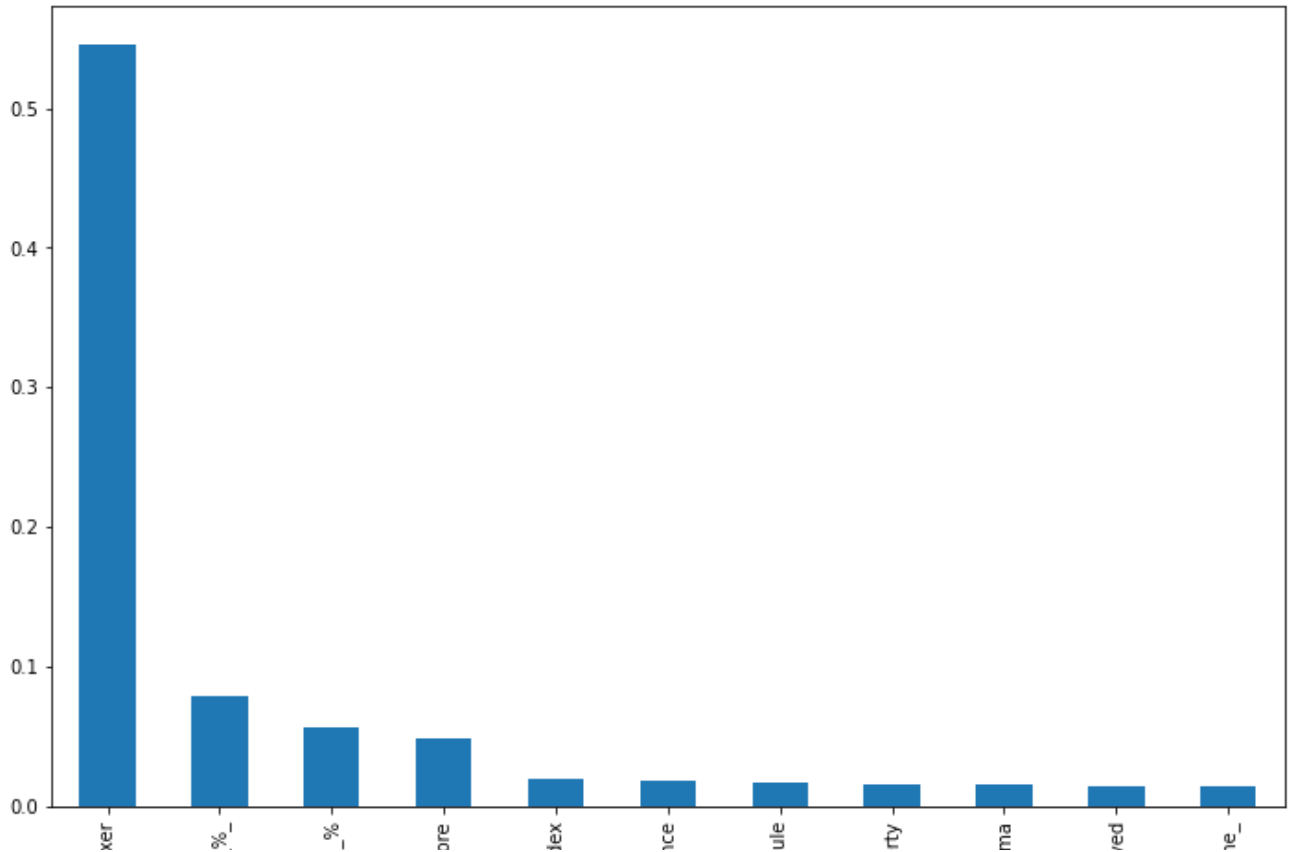
Visualizing the most important features base of the feature importance algorithms with the 10 variables

```
1 plt.figure(figsize=(12,8))
```

```
2 feat_importance.nlargest(11).plot(kind='bar')
```

```
3
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f8bf1b91b70>



- ▼ Here, found the top 10 high correlated columns. Also, let's filter out the columns to report the chosen columns

```
1 indexed=indexed.drop('target')
2 indexed=indexed.withColumnRenamed('target_indexr','target')
3 indexed.show(5)
```

community_area_number	school_id	zip_code	track_schedule	healthy_school_certified	sat
4	609852	60625	0.0	0.0	0.0
10	609937	60656	0.0	0.0	0.0
24	609828	60622	1.0	0.0	0.0
6	610355	60613	0.0	0.0	0.0
61	609929	60609	1.0	0.0	0.0

only showing top 5 rows

```
1 final_data_clean_data=indexed.select(['per_capita_income_',
2                                     'percent_households_below_poverty',
3                                     'percent_aged_16+_unemployed',
4                                     'college_enrollment',
5                                     'percent_aged_25+_without_high_school_diploma',
6                                     'hardship_index',
```



```

7         'average_student_attendance',
8         'safety_score',
9         'isat_exceeding_math_%',
10        'isat_exceeding_reading_%_',
11        'target_indexer']])

```

Here is my final data, clean data, all columns with the high

- ▼ correlation and selection feature process have done with many Machine Learning algorithms.

```
1 final_data_clean_data.show(20)
```

```

+-----+-----+-----+-----+
|per_capita_income_|percent_households_below_poverty|percent_aged_16+_unemployed|coll
+-----+-----+-----+-----+
|          37524|          10.9|          8.2|
|          32875|           5.4|          9.0|
|          43198|          14.7|          6.6|
|          60058|          11.4|          4.7|
|          12765|          29.0|         23.0|
|          44689|          20.6|         10.7|
|          12765|          29.0|         23.0|
|          15528|          27.6|         28.3|
|          12765|          29.0|         23.0|
|          17285|          29.6|         23.0|
|          10402|          30.7|         15.8|
|          35911|          21.7|         15.7|
|          13089|          23.6|         13.9|
|          22677|          17.1|          9.6|
|          43198|          14.7|          6.6|
|          12171|          23.4|         18.2|
|          12961|          42.4|         19.6|
|          57123|           7.5|          5.2|
|          16444|          25.8|         15.8|
|          19398|          31.1|         20.0|
+-----+-----+-----+-----+

```

only showing top 20 rows

- ▼ Our final clean data a Spark engine on top of the Sqlite3 data base.

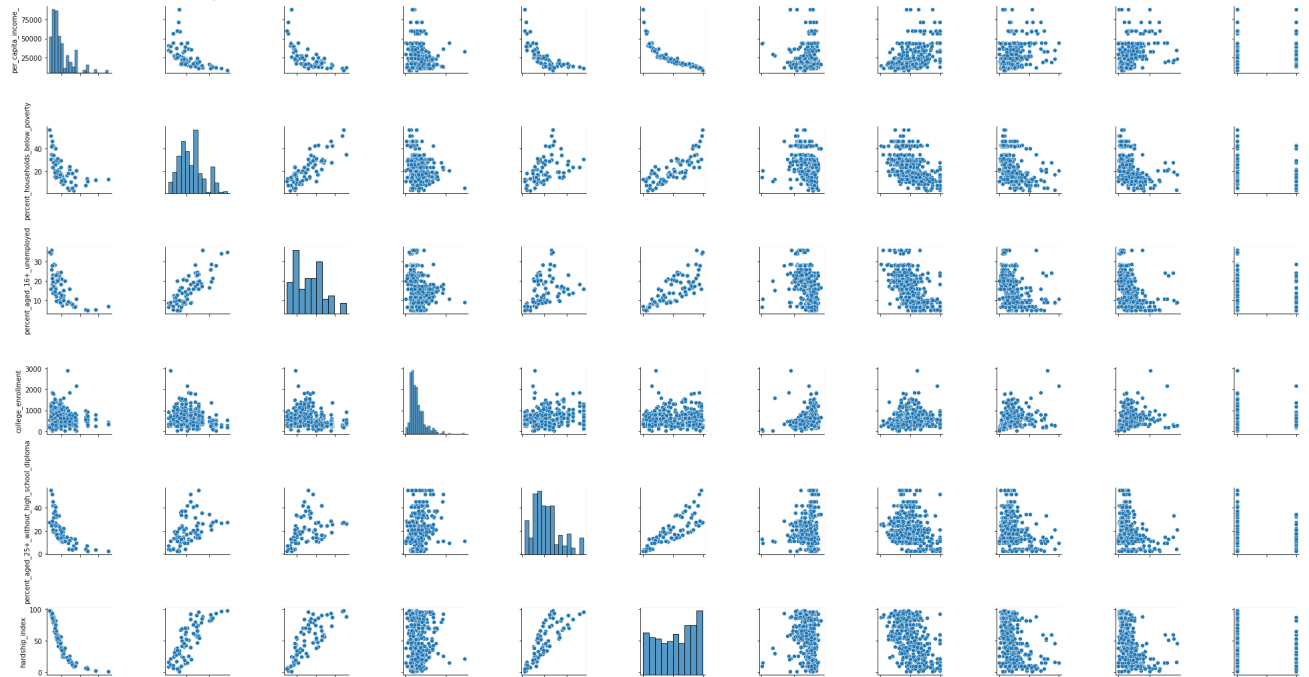
```
1 type(final_data_clean_data)
```

```
pyspark.sql.dataframe.DataFrame
```

Let's get have an idea what are the relation ships amoung all the

```
1 sns.pairplot(final_data_clean_data.toPandas())
```

```
<seaborn.axisgrid.PairGrid at 0x7f8bf1d1ec18>
```



It seems that ISAT math and reading have a highly linearly relation, Also, as it seems there are some skewness and unnormality in the data, so it is better to transform and normalize the data to get the better and precise results for our final goal.

## Modelig the Data



Research question was, did the student do well according the prepared data information and the responses were YES, or NO. Therefore, classification is the one that can be implemented in our case. Further, I want to find a model that responds to the question and it should be a Yes or No response.

## Logestic Regresion

In this project, I am using the Big Data Application Analysis With Spark and SQL, using the Mlib library

The Mlib machine learning libraries just accept two final columns to be processing one is feature and other is a label column, so for that we should transform all the data somehow that the model understands it. Therefore, we should use the Vectorization technique to transform our data to two vectors, which one is vector feature and the other is the label as told, so and in my case is target\_indexr, also for reminding purpose the targeted variable is target\_indexer or as it's original name to "Adequate Yearly Progress Made?"

```
1 from pyspark.ml.feature import VectorAssembler
```

```
1 final_data_clean_data.columns
```

```
['per_capita_income_',
 'percent_households_below_poverty',
 'percent_aged_16+_unemployed',
 'college_enrollment',
 'percent_aged_25+_without_high_school_diploma',
 'hardship_index',
 'average_student_attendance',
 'safety_score',
 'isat_exceeding_math_%',
 'isat_exceeding_reading_%',
 'target_indexer']
```

## ▼ Vectorizing the variables

```
1 vec_assembler=VectorAssembler(inputCols=['per_capita_income_',
2     'percent_households_below_poverty',
3     'percent_aged_16+_unemployed',
4     'college_enrollment',
5     'percent_aged_25+_without_high_school_diploma',
6     'hardship_index',
7     'average_student_attendance',
8     'safety_score',
9     'isat_exceeding_math_%',
10    'isat_exceeding_reading_%'],outputCol='features')
```

```
1 out_vec=vec_assembler.transform(final_data_clean_data.drop('target_indexer'))
```

## Scaling the Data

It is an important task scaling and normalizing the data. We should scale the data before more progress at this point.

```
1 from pyspark.ml.feature import StandardScaler
```

```
1 scaler=StandardScaler(inputCol='features',outputCol='scaledFeatures',verbose=False)
```

```
1 scalerModel=scaler.fit(out_vec)
```

```
1 final_vec=scalerModel.transform(out_vec)
```

```
1 final_vec.show()
```

per_capita_income_	percent_households_below_poverty	percent_aged_16+_unemployed	col1
37524	10.9	8.2	
32875	5.4	9.0	
43198	14.7	6.6	
60058	11.4	4.7	
12765	29.0	23.0	
44689	20.6	10.7	
12765	29.0	23.0	
15528	27.6	28.3	
12765	29.0	23.0	
17285	29.6	23.0	
10402	30.7	15.8	
35911	21.7	15.7	
13089	23.6	13.9	
22677	17.1	9.6	
43198	14.7	6.6	
12171	23.4	18.2	
12961	42.4	19.6	
57123	7.5	5.2	
16444	25.8	15.8	
19398	31.1	20.0	

only showing top 20 rows

```
1 final_vec_selection=final_vec.select('features','target_indexer')
```

```
1 final_vec_selection.show()
```

```
+-----+
|          features|target_indexer|
+-----+
|[37524.0,10.9,8.2...|          1.0|
|[32875.0,5.4,9.0,...|          0.0|
|[43198.0,14.7,6.6...|          1.0|
|[60058.0,11.4,4.7...|          1.0|
|[12765.0,29.0,23....|          0.0|
|[44689.0,20.6,10....|          0.0|
|[12765.0,29.0,23....|          0.0|
|[15528.0,27.6,28....|          0.0|
|[12765.0,29.0,23....|          0.0|
|[17285.0,29.6,23....|          0.0|
|[10402.0,30.7,15....|          0.0|
|[35911.0,21.7,15....|          0.0|
|[13089.0,23.6,13....|          0.0|
|[22677.0,17.1,9.6...|          0.0|
|[43198.0,14.7,6.6...|          0.0|
|[12171.0,23.4,18....|          0.0|
|[12961.0,42.4,19....|          0.0|
|[57123.0,7.5,5.2,...|          0.0|
|[16444.0,25.8,15....|          0.0|
|[19398.0,31.1,20....|          0.0|
+-----+
only showing top 20 rows
```

Training data and testing data is another important procedures that it should be done if we want to have a precise and accurate results.

## ▼ Train Test Split

```
1 train,test=final_vec_selection.randomSplit([0.7,0.3])

1 from pyspark.ml.classification import LogisticRegression
```

## ▼ Fitting the model

```
1 lR=LogisticRegression(labelCol='target_indexer')

1 lRModel=lR.fit(train)
```

## ▼ Model Evaluation

```
1 lrModel.summary.accuracy
```

```
0.9405940594059405
```

▼ Got a 95% accuracy which is awesome, the one reason for this is that the cleaning the data has completed with very caution!

```
1 from pyspark.ml.evaluation import BinaryClassificationEvaluator
```

## ▼ More evaluation on test data

```
1 pred_and_labels=lrModel.evaluate(test)
```

```
1 pred_and_labels.predictions.show()
```

features	target_indexer	rawPrediction	probability	prediction
[60058.0,11.4,4.7...]	1.0	[-0.6515596351473...]	[0.34263816322578...]	0
[12765.0,29.0,23...]	0.0	[7.29700090165460...]	[0.99932289096374...]	1
[12765.0,29.0,23...]	0.0	[6.96074077557897...]	[0.99905250483119...]	1
[15528.0,27.6,28...]	0.0	[4.33107666639969...]	[0.98701738715167...]	1
[44689.0,20.6,10...]	0.0	[4.24415312603299...]	[0.98585507036340...]	1
[10402.0,30.7,15...]	0.0	[6.24319313358492...]	[0.99806013110185...]	1
[17285.0,29.6,23...]	0.0	[4.15767516075531...]	[0.98459707637453...]	1
[35911.0,21.7,15...]	0.0	[3.87049129402733...]	[0.97957764347850...]	1
[13089.0,23.6,13...]	0.0	[5.75033280462228...]	[0.99682836951703...]	1
[12171.0,23.4,18...]	0.0	[7.57340917750129...]	[0.99948632664099...]	1
[15957.0,28.6,22...]	0.0	[4.76899566805198...]	[0.99158255332108...]	1
[57123.0,7.5,5.2,...]	0.0	[2.47310912392501...]	[0.92223503662239...]	0
[17104.0,19.2,12...]	0.0	[5.03173997964759...]	[0.99351488813276...]	1
[44689.0,20.6,10...]	0.0	[3.46789338736027...]	[0.96976030265187...]	1
[15957.0,28.6,22...]	0.0	[4.65861776877529...]	[0.99060946184607...]	1
[23482.0,10.4,11...]	0.0	[4.33040866835338...]	[0.98700882459623...]	1
[43198.0,14.7,6.6...]	1.0	[-1.2483477625146...]	[0.22298628007815...]	0
[12034.0,43.1,21...]	0.0	[4.95375309879908...]	[0.99299257648424...]	1
[23791.0,29.6,18...]	0.0	[2.29927345140624...]	[0.90881684866717...]	0
[16563.0,25.9,19...]	0.0	[4.53935387111489...]	[0.98943255832379...]	1

only showing top 20 rows

Comparing the predictions and the original values shows that how

▼ AUC -----> area under the curve evaluation process

```
1 AUC_eval=BinaryClassificationEvaluator(rawPredictionCol='prediction',]
```

```
1 AUC=AUC_eval.evaluate(pred_and_labels.predictions)
```

▼ 87% accuracy with the AUC test which is pretty good.

```
1 AUC
```

```
0.7898073022312374
```

```
1 final_data_clean_data.show(6)
```

```
+-----+-----+-----+-----+
|per_capita_income_|percent_households_below_poverty|percent_aged_16+_unemployed|coll|
+-----+-----+-----+-----+
|          37524|          10.9|          8.2|
|          32875|           5.4|          9.0|
|          43198|          14.7|          6.6|
|          60058|          11.4|          4.7|
|          12765|          29.0|         23.0|
|          44689|          20.6|         10.7|
+-----+-----+-----+-----+
```

only showing top 6 rows

## Reports to Census

With Quering and analyzing the data based on the machin learning techniques and feature selecting techniques, it has found that there are some columns to be consisder from amoung of the data sets.

1.per\_capita\_income\_,



- 2.percent\_households\_below\_poverty,
- 3.percent\_aged\_16+\_unemployed,
- 4.college\_enrollment,
- 5.percent\_aged\_25+\_without\_high\_school\_diploma,
- 6.hardship\_index,
- 7.average\_student\_attendance,
- 8.safety\_score,
- 9.isat\_exceeding\_math\_%,
- 10.isat\_exceeding\_reading\_%\_

These are the most highly coefficiently linearly related and as a results these coloumns could be considered for further decisions.Also these predictors are trying to respond to the target variable as it is in the original data as "Adequate Yearly Progress Made?"

