

Predict Students' Dropout and Academic Success using Classification Models

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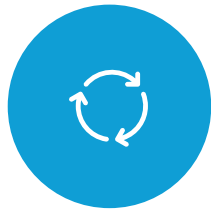
Motivation



COUNTRY GDP IS HIGHLY
INFLUENCED BY THE
LABOR FORCE



ESPECIALLY SKILLED
LABOR FORCE



COUNTRY ECONOMIC
CYCLE IS CAN BE
DESCRIBED BY AD-AS
DIAGRAM.



AGGREGATE DEMAND
AND AGGREGATE SUPPLY



AGGREGATE SUPPLY IS
THE TOTAL AMOUNT OF
GOODS (INCLUDING
SERVICES) SUPPLIED BY
BUSINESSES WITHIN A
COUNTRY AT A GIVEN
PRICE LEVEL.
([HTTPS://WWW.ECONLIB.
ORG/](https://www.econlib.org/))



LAND , LABOR, CAPITAL,
TECHNOLOGY

Motivation



Changes in Education and Skills



Impact: A more educated and skilled workforce can produce more output and adapt to new technologies, enhancing long-run AS.



Examples:

Investments in STEM (Science, Technology, Engineering, and Mathematics) education leading to a more innovative workforce in countries like South Korea.

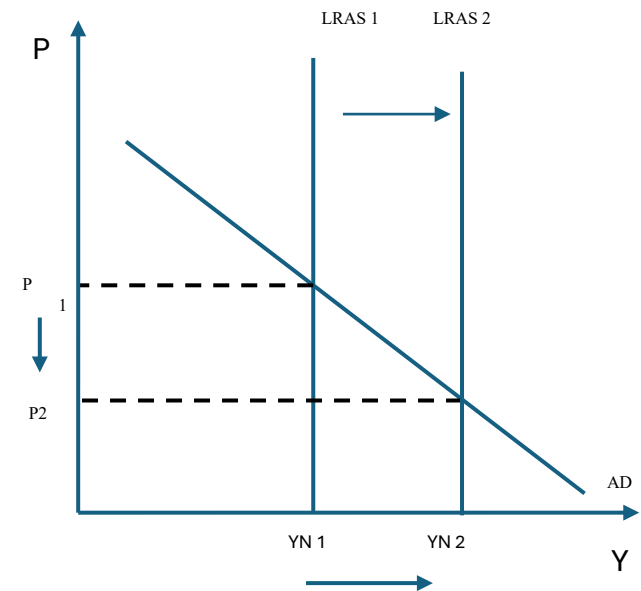
Vocational training programs in Germany that equip workers with specialized skills, increasing productivity.



University and colleges are places where skilled labor are produced.

How?

When skilled labor increases in the labor market, output in the market will rise, pushing the long-run aggregate supply (LRAS) to the right. This causes the long-term output level to increase, leading to higher GDP and a fall in the price level, which is a positive sign for the economy.



Dataset



Predict Students' Dropout and Academic Success

Donated on 12/12/2021

A dataset created from a higher education institution (acquired from several disjoint databases) related to students enrolled in different undergraduate degrees, such as agronomy, design, education, nursing, journalism, management, social service, and...



Dataset Characteristics

Tabular

Subject Area

Social Science

Associated Tasks

Classification

Feature Type

Real, Categorical, Integer

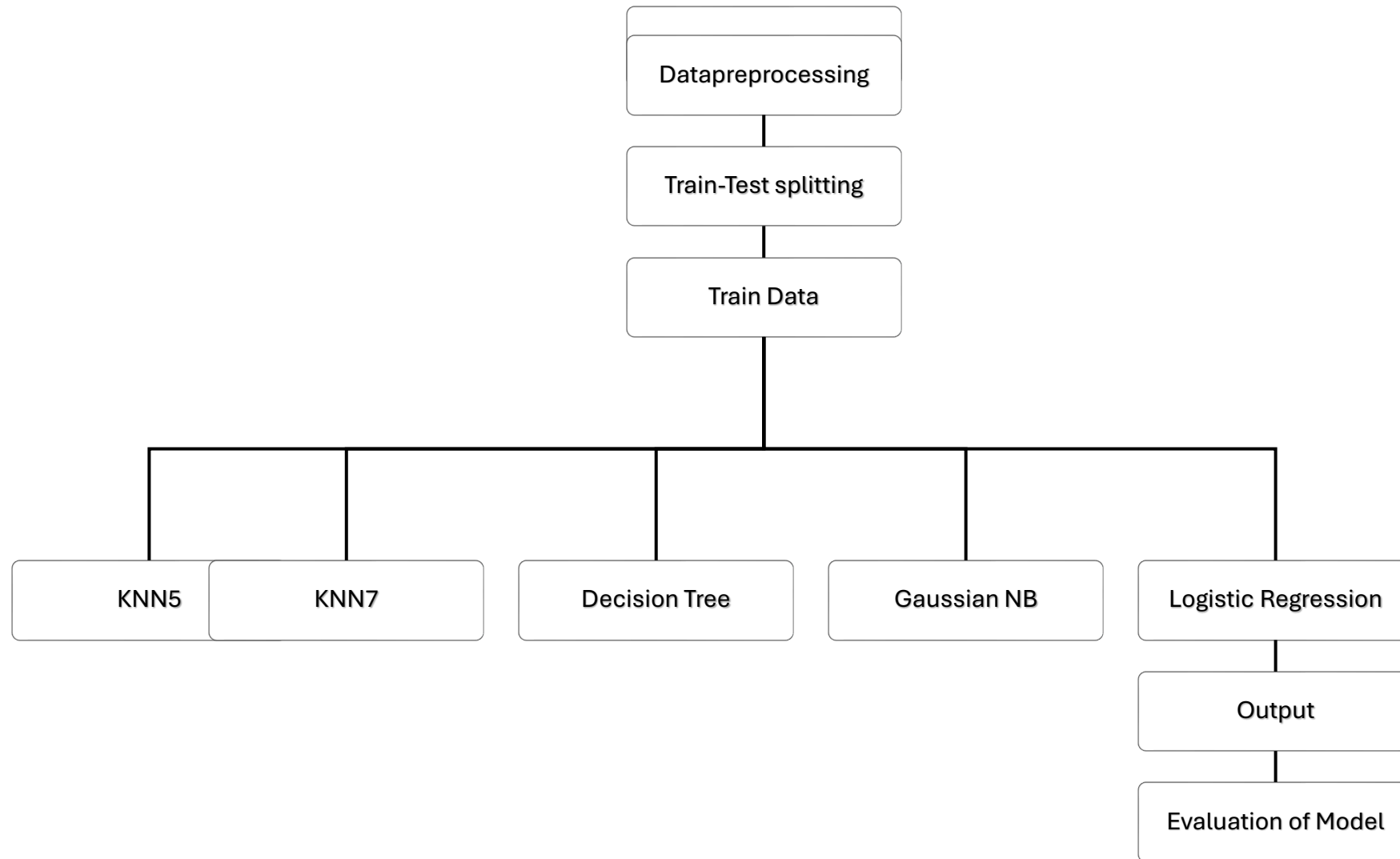
Instances

4424

Features

36

Method



Target value

```
[84] data['Target'].value_counts()
```



count

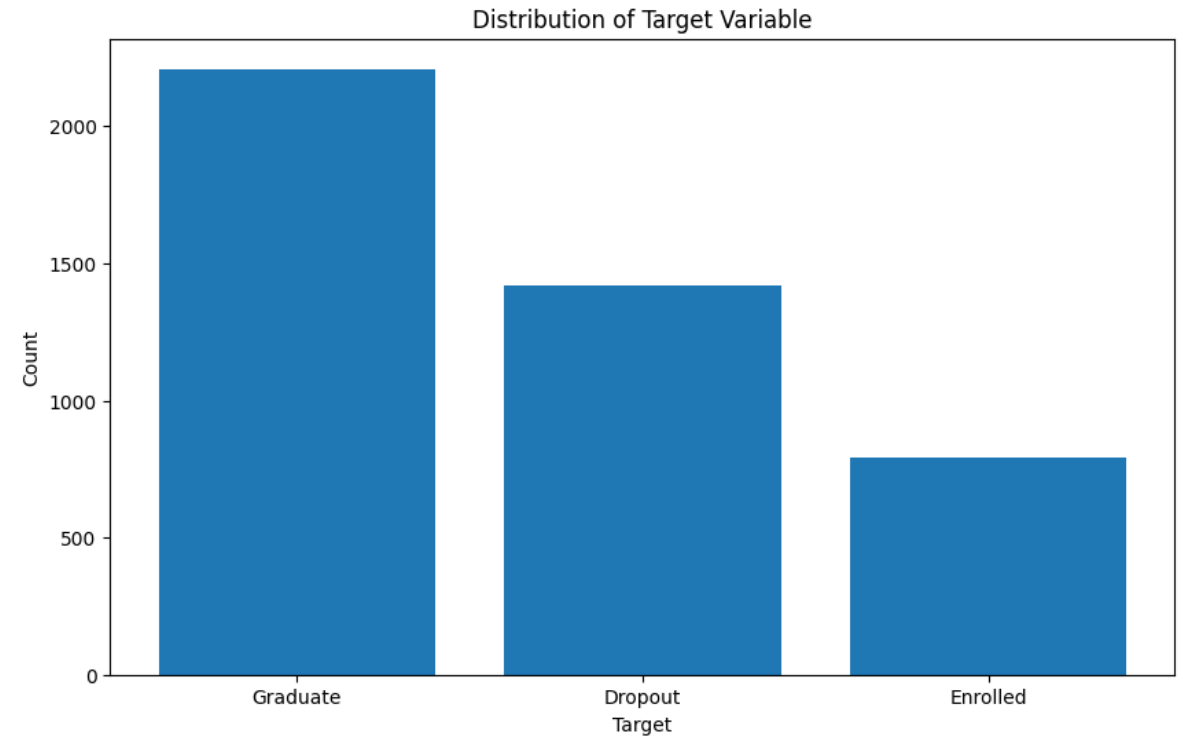
Target

Graduate 2209

Dropout 1421

Enrolled 794

dtype: int64



Drop the “enrolled” data

```
data = data[data.Target != 'Enrolled']  
data['Target'].value_counts()
```

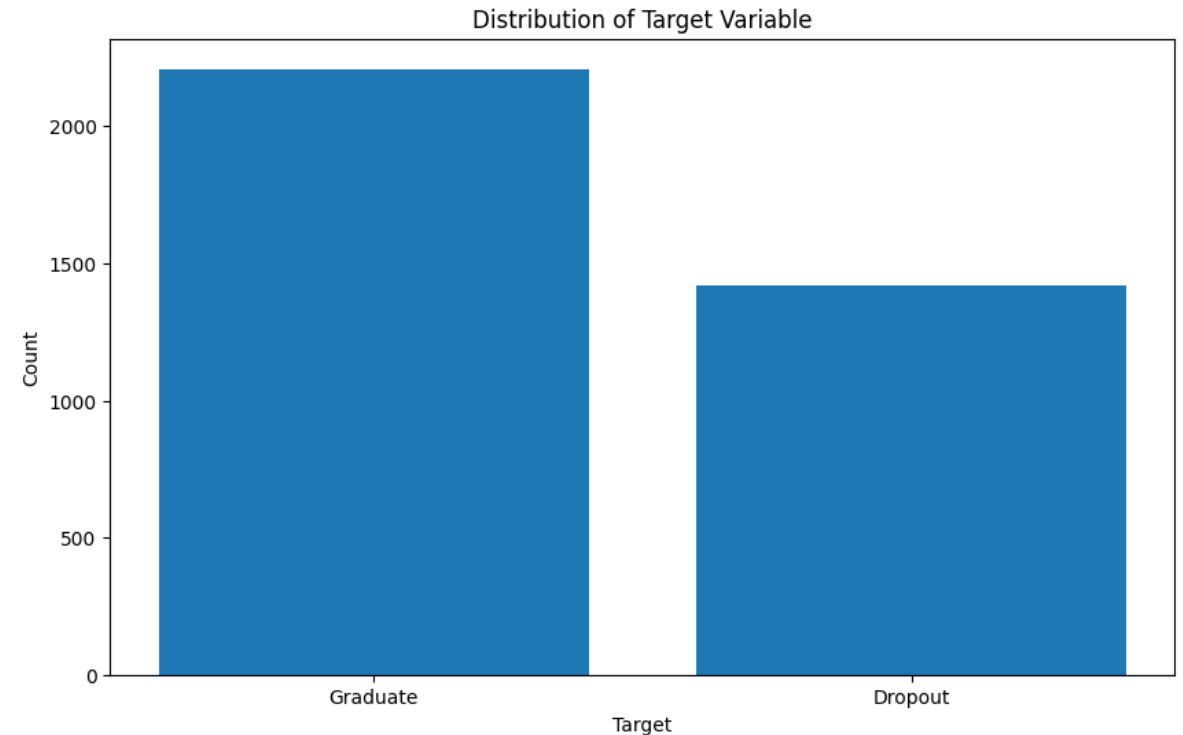
count

Target

Graduate 2209

Dropout 1421

dtype: int64




```
#encode the target  
data['Target1'] = LabelEncoder().fit_transform(data['Target'])  
data['Target1'].value_counts()
```

count

Target1	count
1	2209
0	1421

dtype: int64

```
[ ] data['Target1'] = data['Target1'].apply(lambda x: 1 if x == 0 else 0 )  
data['Target1'].value_counts()
```

count

Target1	count
0	2209
1	1421

dtype: int64

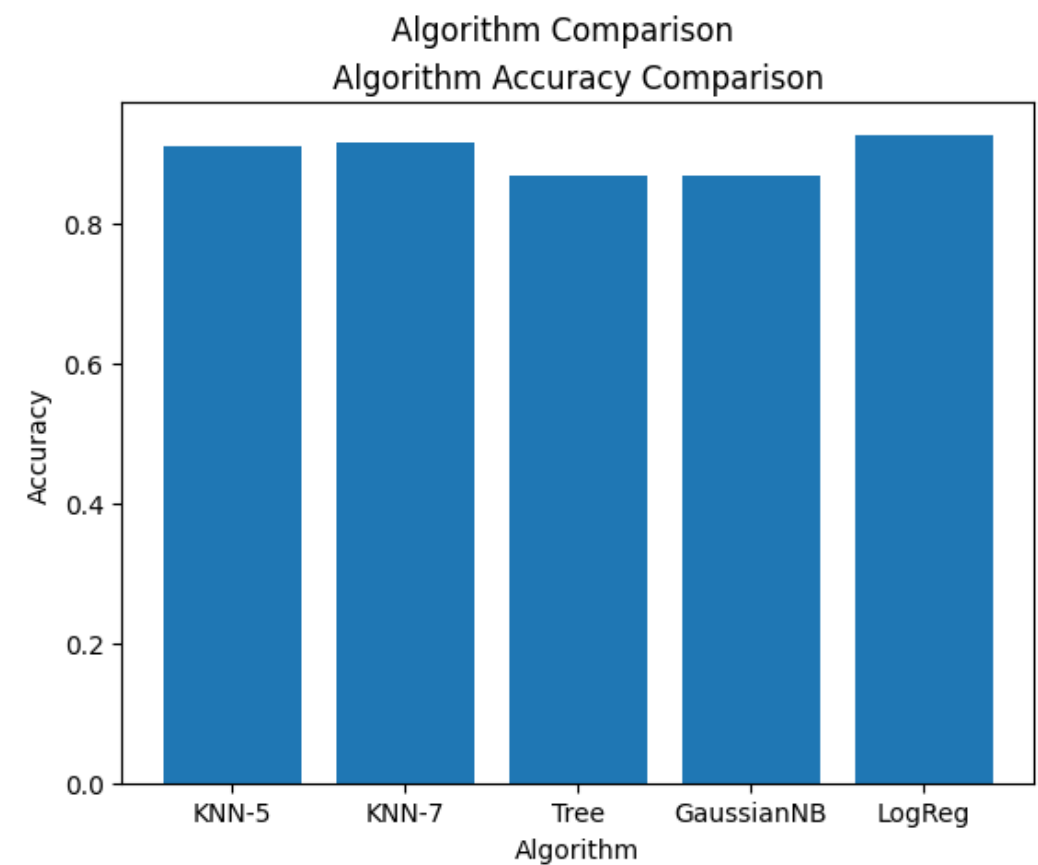
Encoding the data

Correlation

	Target
Target	1.000000
Curricular units 2nd sem (approved)	0.624157
Curricular units 2nd sem (grade)	0.566827
Curricular units 1st sem (approved)	0.529123
Curricular units 1st sem (grade)	0.485207
Tuition fees up to date	0.409827
Scholarship holder	0.297595
Curricular units 2nd sem (enrolled)	0.175847
Curricular units 1st sem (enrolled)	0.155974
Admission grade	0.120889
Displaced	0.113986
Previous qualification (grade)	0.103764

```
x = data[["Previous qualification (grade)",  
         "Displaced",  
         "Admission grade",  
         "Curricular units 1st sem (enrolled)",  
         "Curricular units 2nd sem (enrolled)",  
         "Scholarship holder",  
         "Tuition fees up to date",  
         "Curricular units 1st sem (approved)",  
         "Curricular units 1st sem (grade)",  
         "Curricular units 2nd sem (approved)",  
         "Curricular units 2nd sem (grade)"]].values
```

Results and Evaluation



Model	Accuracy	Precision	Recall	F1-Score
KNN-5	0.910468	0.926070	0.838028	0.879852
KNN-7	0.915978	0.930502	0.848592	0.887661
Tree	0.867769	0.815436	0.855634	0.835052
GaussianNB	0.867769	0.894958	0.750000	0.816092
LogReg	0.926997	0.913978	0.897887	0.905861



Question ?