Data Set Information:

This data approach student achievement in secondary education of two Portuguese schools. The data attributes include student grades, demographic, social and school related features) and it was collected by using school reports and questionnaires. Two datasets are provided regarding the performance in two distinct subjects: Mathematics (mat) and Portuguese language (por). In [Cortez and Silva, 2008], the two datasets were modeled under binary/five-level classification and regression tasks. Important note: the target attribute G3 has a strong correlation with attributes G2 and G1. This occurs because G3 is the final year grade (issued at the 3rd period), while G1 and G2 correspond to the 1st and 2nd period grades. It is more difficult to predict G3 without G2 and G1, but such prediction is much more useful.

Attribute Information:

# Attributes for both student-mat.csv (Math course) and student-por.csv (Portuguese language course) datasets:

1 school - student's school (binary: 'GP' - Gabriel Pereira or 'MS' - Mousinho da Silveira)

2 sex - student's sex (binary: 'F' - female or 'M' - male)

3 age - student's age (numeric: from 15 to 22)

4 address - student's home address type (binary: 'U' - urban or 'R' - rural)

5 famsize - family size (binary: 'LE3' - less or equal to 3 or 'GT3' - greater than 3)

6 Pstatus - parent's cohabitation status (binary: 'T' - living together or 'A' - apart)

7 Medu - mother's education (numeric: 0 - none, 1 - primary education (4th grade), 2 â€“ 5th to 9th grade, 3 â€“ secondary education or 4 â€“ higher education)

8 Fedu - father's education (numeric: 0 - none, 1 - primary education (4th grade), 2 â€“ 5th to 9th grade, 3 â€“ secondary education or 4 â€“ higher education)

9 Mjob - mother's job (nominal: 'teacher', 'health' care related, civil 'services' (e.g. administrative or police), 'at\_home' or 'other')

10 Fjob - father's job (nominal: 'teacher', 'health' care related, civil 'services' (e.g. administrative or police), 'at\_home' or 'other')

11 reason - reason to choose this school (nominal: close to 'home', school 'reputation', 'course' preference or 'other')

12 guardian - student's guardian (nominal: 'mother', 'father' or 'other')

13 traveltime - home to school travel time (numeric: 1 - <15 min., 2 - 15 to 30 min., 3 - 30 min. to 1 hour, or 4 - >1 hour)

14 studytime - weekly study time (numeric: 1 - <2 hours, 2 - 2 to 5 hours, 3 - 5 to 10 hours, or 4 - >10 hours)

15 failures - number of past class failures (numeric: n if 1<=n<3, else 4)

16 schoolsup - extra educational support (binary: yes or no)

17 famsup - family educational support (binary: yes or no)

18 paid - extra paid classes within the course subject (Math or Portuguese) (binary: yes or no)

19 activities - extra-curricular activities (binary: yes or no)

20 nursery - attended nursery school (binary: yes or no)

21 higher - wants to take higher education (binary: yes or no)

22 internet - Internet access at home (binary: yes or no)

23 romantic - with a romantic relationship (binary: yes or no)

24 famrel - quality of family relationships (numeric: from 1 - very bad to 5 - excellent)

25 freetime - free time after school (numeric: from 1 - very low to 5 - very high)

26 goout - going out with friends (numeric: from 1 - very low to 5 - very high)

27 Dalc - workday alcohol consumption (numeric: from 1 - very low to 5 - very high)

28 Walc - weekend alcohol consumption (numeric: from 1 - very low to 5 - very high)

29 health - current health status (numeric: from 1 - very bad to 5 - very good)

30 absences - number of school absences (numeric: from 0 to 93)

# these grades are related with the course subject, Math or Portuguese:

31 G1 - first period grade (numeric: from 0 to 20)

31 G2 - second period grade (numeric: from 0 to 20)

32 G3 - final grade (numeric: from 0 to 20, output target)

* Therefore we have to create a model without G1 and G2.
* Then we can category G3 value into 3 category as,
  + 0% =< poor < 45%
  + 45% =< average < 75%
  + 75% =< good < 100%
* Data set is imbalance, because G3 value categories,
  + B 692
  + C 230
  + A 122
* The challenge of working with imbalanced datasets is that most machine learning techniques will ignore, and in turn have poor performance on, the minority class, although typically it is performance on the minority class that is most important. To solve this we use Synthetic Minority Oversampling Technique or SMOTE. The basic explanation of SMOTE is that it will overpopulate the minority class by synthesizing new examples by selecting those that are close to its feature space. Or simply, it will generate new rows for the class that is under-represented in our dataset. [[ref]](https://towardsdatascience.com/evaluation-and-prediction-of-students-academic-performance-during-covid-19-40bb2b90141b)
* In our Decision tree classifier most important 10 features are,
  + absences
  + failures
  + freetime
  + higher
  + Medu
  + paid
  + Walc
  + goout
  + reason
  + age
* Changing the maxdepth of the model tried to find best model.
* Then check the overfitting and underfitting by checking training and testing accuracies
* Then try to show the accuracy best on the k-fold cross validation
  + When k=05 accuracy = 0.701
  + When k=10 accuracy = 0.725
* Then tried with Random Forest Classifier and most important 10 features are,
  + absences
  + Medu
  + failures
  + goout
  + Walc
  + health
  + freetime
  + Fedu
  + reason
  + paid
* Changing the maxdepth of the model tried to find best model.
* Then check the overfitting and underfitting by checking training and testing accuracies
* Then try to show the accuracy best on the k-fold cross validation
  + When k=05 accuracy = 0.792
  + When k=10 accuracy = 0.796
* Then try SVM with [OneVsOneClassifier](https://scikit-learn.org/stable/modules/generated/sklearn.multiclass.OneVsOneClassifier.html)
* Then check the overfitting and underfitting by checking training and testing accuracies
* Then try to show the accuracy best on the k-fold cross validation
  + When k=05 accuracy = 0.703
  + When k=10 accuracy = 0.706
* Then try KNN
* Changing the maxdepth of the model tried to find best model.
* Then check the overfitting and underfitting by checking training and testing accuracies
* Then try to show the accuracy best on the k-fold cross validation
  + When k=05 accuracy = 0.797
  + When k=10 accuracy = 0.802