Prediction of Academic Performance of College Students with Bipolar Disorder using different Deep learning and Machine learning algorithms

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Abstract

In modern years, the performance of the students is analysed with lot of difficulties, which is a very important problem in all the academic institutions. The main idea of this paper is to analyze and evaluate the academic performance of the college students with bipolar disorder by applying data mining classification algorithms using Jupiter Notebook, python tool. This tool has been generally used as a decision-making tool in terms of academic performance of the students. The various classifiers could be logistic regression, random forest classifier gini, random forest classifier entropy, decision tree classifier, K-Neighbours classifier, Ada Boost classifier, Extra Tree Classifier, GaussianNB, BernoulliNB are used. The results of such classification model deals with 13 measures like Accuracy, Precision, Recall, F1 Measure, Sensitivity, Specificity, R Squared, Mean Absolute Error, Mean Squared Error, Root Mean Squared Error, TPR, TNR, FPR and FNR. Therefore, conclusion could be reached that the Decision Tree Classifier is better than that of different algorithms.

Keywords: Logistic Regression, Random Forest classifier, Decision tree classifier, K-Neighbours classifier, Ada Boost classifier, Extra Tree classifier, GaussianNB.

I. INTRODUCTION

Data Mining could be a promising and prospering wilderness in the examination of information and moreover, the outcome of the investigation has numerous applications. Information Mining can likewise allude to Knowledge Discovery from Data (KDD). This framework capacities as the machine-driven or helpful extraction of examples addressing information verifiably keep or caught in immense data sets, information stockrooms, the Web, information archives, and data streams. Data Mining is a multidisciplinary field, enveloping regions like data innovation, AI, insights, design acknowledgment, information recovery, neural organizations, data-based frameworks, man-made reasoning, and information perception.

The use of information mining is generally common in training frameworks. Instructive information mining is an

arising field that can be successfully applied in the field of training. The instructive information mining employments a few thoughts and ideas, for example, Association rule mining, grouping, and bunching. The information that arises can be utilized to all the more likely comprehend student's' advancement rate, student's' consistency standard, student's' progress rate, and the student's' prosperity. The information mining framework is urgent and essential to gauge the student's' execution improvement. The order calculations can be utilized to characterize and investigate the student's' information set in precise way. The attributes that may be used for student's academic performance such Name, College, Class Tutor Name, Gender, Age, Address, Family size, Parent's Status, Mother's Education, Father's Education, Mother's Job, Father's Job, Reason to choose this college, Student's Guardian, Travel Time, Study time (Weekly), Number of Subjects failed so far, College Support (Extra curricular), Family Educational Support, Paid courses attended, Extra curricular Activities involves, Nursery schools attended, Higher education, Internet access at home, love relationships, Family relationships, Free time after college, Going out with friends, Week days - Alcohol / Smoking, Weekends -Alcohol / Smoking, Current Health Status, Number of days absent, CGPA, Number of Psychological motivation sessions attended, CGPA Grade.

The main idea of this research paper is to use data mining classification algorithms to study and analyze the academic performance of the college students with bipolar disorder. This research paper constitutes 5 sections. Section-1 deals with introduction; Section-2 enumerates a related work. Section-3 presents the idea and aspects of different classifiers. Section-4 elaborates the Data Preprocessing. Section-5 explains the implementation of model construction. Section-6 handles results and discussions. Conclusion will be given in Section-7.

II. LITERATUR SURVEY

Academic performance of students in advanced education (HE) is investigated broadly to handle scholarly underachievement, expanded college dropout rates, graduation delays, among other persevering difficulties [1]. In straightforward terms, student performance alludes to the degree of accomplishing present moment and long haul objectives in education [2]. Nonetheless, academicians measure student's accomplishment according to alternate points of view, going from student's' last grades, grade point normal (GPA), to future occupation possibilities [3]. The writing offers an abundance of computational efforts striving to further develop student execution in schools and colleges, most remarkably those determined by data mining and learning investigation procedures [4]. The opportune expectation of student execution empowers identification of low performing student's, along these lines, engaging instructors to mediate right on time during the learning cycle and carry out the necessary intercessions. Productive intercessions incorporate, however are not restricted to, student prompting, execution progress observing, shrewd mentoring frameworks advancement, and policymaking [5]. This undertaking is emphatically supported by computational advances in information mining and learning examination [6]. A new complete study features that around 70% of the checked on work explored student execution expectation utilizing student grades and GPAs, while just 10% of the investigations reviewed the forecast of student accomplishment utilizing learning results [3]. This hole impelled us to altogether research the work did where the learning results are utilized as an intermediary for student scholastic execution. Result based instruction is a worldview of schooling that spotlights on executing and achieving the purported learning results [7]. As a result, student learning results are objectives that action the degree which student's achieves the proposed capabilities, explicitly information, abilities, and qualities, toward the finish of a specific learning measure. In our view, the student results address a more comprehensive measurement for making a decision about student scholastic accomplishments than simple appraisal grades. This view agrees with the case that the learning results address basic components of student scholarly achievement [8]. Also, eminent HE accreditation associations, like ABET and ACBSP, utilize the learning results as the structure blocks for surveying the nature of instructive projects [9]. Such significance calls for more exploration endeavours to anticipate the fulfilment of learning results, both at the course and program levels. The absence of methodical reviews exploring the expectation of student execution utilizing student results has roused us to seek after the destinations of this exploration. In an efficient writing survey (i.e., SLR), a bit by bit convention is executed to

distinguish, select, and assess the integrated investigations to respond to explicit examination questions [10, 11].

III. CLASSIFICATION

Classification is a process that is used to categorize data into predefined unconditional class labels. Classification can be a two venture process comprising of training and testing. In the initial step, a model is developed by investigating the data tuples from training data having a collection of attributes. For each tuple in the preparing data, the value of class label attribute is understood. Categorization rule is applied on preparing data to frame the model. In the second step of arrangement, test data is utilized to look at the exactness of the model. Assuming the precision of the model is proper, the model can be utilized to group the unknown information tuples.

3.1. Classification Algorithms

The fundamental techniques for classification used in this work are logistic regression, random forest classifier gini, random forest classifier entropy, decision tree classifier, K-Neighbours classifier, Ada Boost classifier, Extra Tree Classifier, GaussianNB, and BernoulliNB. We will discuss those classifiers one by one.

3.2 Logistic regression

Logistic regression is a supervised learning classification algorithm used to predict the probability of a target variable. The nature of target or dependent variable is dichotomous, which means there would be only two possible classes.

3.3 Random Forest Classifier gini

Random Forest Classifier gini or Mean Decrease in Impurity (MDI) calculates each feature importance as the sum over the number of splits (across all tress) that include the feature, proportionally to the number of samples it splits.

3.4 Random Forest Classifier Entropy

Random Forest Classifier Entropy helps us to select the best splitter. It can be defined as the measure of the purity of the sub split. It always lies between 0 and 1.

$$H(s) = -P(+) log_2P(+) - P(-) log_2P(-)$$

3.5 Decision Tree Classifier

Decision tree develops classification or regression models as a tree structure. It's everything except a dataset into more humble and more modest subsets while at the same time, a connected decision tree is steadily evolved. The possible outcome is a tree with choice hubs and leaf hubs.

3.6 K-Neighbours Classifier

K neighbours classifier is used to store all available cases and classifies new cased based on similarity measure such as distance functions. It has been used in statistical estimation and pattern recognition.

3.7 Ada Booster Classifier

It is a boosting procedure that is utilized as a gathering technique in AI, in which the weights are re-allotted to each case, with higher loads to inaccurately classified cases.

3.8 Extra Tree Classifier

It is a kind of outfit learning procedure that totals the results of various de-associated choice trees gathered in a "forest" to yield its grouping result.

3.9 GaussianNB Classifier

It is the fast, accurate and reliable algorithm that has high accuracy and speed on large datasets. It assumes that the effect of a particular feature in a class in independent of other features.

3.10 BernoulliNB Classifier

It is designed for binary / Boolean features. Threshold for binarizing (mapping to Booleans) of sample features.

IV. DATA PRE-PROCESSING

Datasets used inside classification algorithm should be clear and can be pre-processed for taking care of absent or excess attributes. The information are to be taken care of with effectiveness to prompt the best result from the Data Mining measure.

4.1 Identification of Attributes

Dataset collected from college students database are given with the attributes.

Table-1 Symbolic Attribute Description

| Attributes | Description | Possible Values |
|------------|---------------------------|---|
| Name | Name of the Student | Text |
| College | Name of the College | Jamal Mohamed College, Holy Cross College, Bishop Heber College, St. Joseph's College, EVR College, National College, Cauvery College for Women, Indira Gandhi College, SRC, MIET, Others |
| CTN | Name of the Class teacher | Text |
| Gender | Gender of the student | Male, Female, Transgender |

| A ga | Age of the students | 18 20 | | | |
|-------------------|---|--|--|--|--|
| Age Address | Age of the students Whether the students | 18 – 30 Urban, City | | | |
| 1 1dd1 C55 | is from city or village | Citan, City | | | |
| FS | Family size | Less than 3, Greater | | | |
| 1.5 | Tunning Size | than 3 | | | |
| PS | To identify whether | Joined, Apart | | | |
| | they are joined or not | , , , , , , , , , , , , , , , , , , , | | | |
| ME | Educational | 0 (None), 1 | | | |
| | qualification of | (Primary Education, | | | |
| | mother | Upto 4th Standard), | | | |
| | | 2 (5 th std to 9 th std), | | | |
| | | 3 (10 th Std to 12 th | | | |
| | | std), 4 (UG / PG) | | | |
| FE | Educational | 0 (None), 1 | | | |
| | qualification of father | (Primary Education, | | | |
| | | Upto 4th Standard), | | | |
| | | $2 (5^{th} \text{ std to } 9^{th} \text{ std}),$ | | | |
| | | 3 (10 th Std to 12 th | | | |
| | | std), 4 (UG / PG) | | | |
| MJ | Job of the mother | Teacher, Health | | | |
| | | care related, Job | | | |
| | | less, Others | | | |
| FJ | Job of the father | Teacher, Health | | | |
| | | care related, Job | | | |
| DTCC | Б | less, Others | | | |
| RTCC | Exact reason for | Close to home, | | | |
| | joining the college | College Reputation, | | | |
| | | Course Preference, Others | | | |
| CC | Guardian of the | | | | |
| SG | student | Father, Mother, Others | | | |
| TT | Time of the travelling | Less than 15 | | | |
| 11 | Time of the travelling | minutes, 15 minutes | | | |
| | | - 30 minutes, 30 | | | |
| | | minutes – 1hour, | | | |
| | | more than 1 hour | | | |
| ST | Time duration of the | Less than 2 hours, 2 | | | |
| | week | - 5 hrs, 5 - 10 hrs, | | | |
| | | more than 10 hrs | | | |
| NSFS | Arrears so far in the | Nil, 1,2, 3 and more | | | |
| | subjects studied | than 3 | | | |
| CSIEC | Importance given by | Yes, No | | | |
| | the college for extra | | | | |
| | curricular activities | | | | |
| FES | Educational Support | Yes, No | | | |
| D.C.I. | given by the family | 77 37 | | | |
| PCA | Any additional course | Yes, No | | | |
| ECA | for cost attended | V N. | | | |
| ECA | Involved in any extra- | Yes, No | | | |
| | curricular activities or | | | | |
| Nurcary | not Attended in Nursery | Yes, No | | | |
| Nursery | | i es, ino | | | |
| HE | school or no | V. N. | | | |
| TIE | Want to join Higher | | | | |
| | Want to join Higher | Yes, No | | | |
| IAH | Education | | | | |
| IAH | Education To identify whether | Yes, No | | | |
| IAH | Education To identify whether the internet is | | | | |
| | Education To identify whether the internet is accessed or not | Yes, No | | | |
| AYL | Education To identify whether the internet is accessed or not Love relationships | Yes, No Yes, No | | | |
| | Education To identify whether the internet is accessed or not Love relationships About relationships of | Yes, No Yes, No Very bad, bad, | | | |
| AYL | Education To identify whether the internet is accessed or not Love relationships | Yes, No Yes, No Very bad, bad, good, very good, | | | |
| AYL | Education To identify whether the internet is accessed or not Love relationships About relationships of | Yes, No Yes, No Very bad, bad, good, very good, excellent | | | |
| AYL FR | Education To identify whether the internet is accessed or not Love relationships About relationships of the family | Yes, No Yes, No Very bad, bad, good, very good, excellent Very bad, bad, | | | |
| AYL FR | Education To identify whether the internet is accessed or not Love relationships About relationships of the family Free time after the | Yes, No Yes, No Very bad, bad, good, very good, excellent | | | |
| AYL FR | Education To identify whether the internet is accessed or not Love relationships About relationships of the family Free time after the class | Yes, No Yes, No Very bad, bad, good, very good, excellent Very bad, bad, good, very good, yery good, very good, | | | |
| AYL FR FTAC | Education To identify whether the internet is accessed or not Love relationships About relationships of the family Free time after the class | Yes, No Yes, No Very bad, bad, good, very good, excellent Very bad, bad, good, very good, excellent | | | |

| WDAC | Alcohol consumption | Very bad, bad, | | | |
|-------|------------------------|--------------------|--|--|--|
| | during Monday to | good, very good, | | | |
| | Friday | excellent | | | |
| WEAC | Alcohol consumption | Very bad, bad, | | | |
| | during Saturday and | good, very good, | | | |
| | Sunday | excellent | | | |
| CHS | Status of the health | Very bad, bad, | | | |
| | | good, very good, | | | |
| | | excellent | | | |
| NDA | Total number of days | 0 to 90 | | | |
| | absent for the classes | | | | |
| CGPA | Cumulative Grade | 0.0 to 10.0 | | | |
| | Point Average | | | | |
| NPMSA | Number of | 0 to 10 | | | |
| | Psychological | | | | |
| | motivation sessions | | | | |
| | attended | | | | |
| CGPAG | Grade for the CGPA | Distinction, First | | | |
| | | Class, Second | | | |
| | | Class, Pass, Fail | | | |

V. IMPLEMENTATION USING JUPITER NOTEBOOK, PYTHON

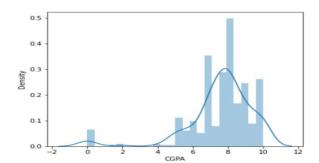
Jupiter Notebook is an open-source web application that permits us to create and share documents that contain real code, equations, and visualizations, and summarized text. It is used for data cleaning, transformation, numerical simulation, statistical modeling, data visualization, machine learning, etc. From the above attributes given, the student analysis.arff file was created and uploaded into Jupiter Notebook. The academic performance of the students is influenced by the above-specified attributes. The various classifiers such as logistic regression, random forest classifier Gini, random forest classifier entropy, decision tree classifier, K-Neighbours classifier, Ada Boost classifier, Extra Tree Classifier, GaussianNB, and BernoulliNB are used. The results of such classification model deal with 13 measures like Accuracy, Precision, Recall, F1 Measure, Sensitivity, Specificity, R Squared, Mean Absolute Error, Mean Squared Error, Root Mean Squared Error, TPR, TNR, FPR, and FNR.

VI. RESULTS AND DISCUSSION

Table 2 Extraction of required features from the dataset

| Address | Study_h ours | Extra_curri culum | Internet_a ccess | Family_inter action | Drug_use_we ekdays | Drug_use_we ekends | Leave _log | CG PA | Psycholo gical motivati on sessions attended | CGPA_G rade |
|--------------------|-----------------|----------------------|---------------------|------------------------|-----------------------|-----------------------|---------------|----------|---|-----------------|
| Rural(Vil lage) | 2 | 0 | 1 | 5 | 1 | 1 | 5.0 | 8.0 | 7 | First Class |
| Rural (Village) | 10 | 0 | 0 | 5 | 1 | 1 | 0.0 | 10.0 | 10 | Outstandi ng |
| Rural (Village) | 2 | 0 | 1 | 5 | 1 | 1 | 5.0 | 9.9 | 10 | Outstandi ng |
| Rural (Village) | 2 | 1 | 1 | 5 | 2 | 2 | 6.0 | 7.0 | 5 | First class |
| Urban (City) | 10 | 1 | 1 | 3 | 1 | 1 | 4.0 | 8.0 | 7 | First class |

We have given the results of CGPA Grade, which is based on psychological motivation sessions demonstrated that the rural students give outstanding performance than urban students.



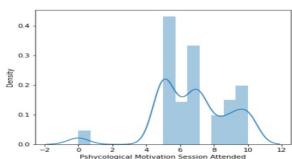


Fig 1: Analysis of CGPA with respect to Psychological motivation session attended

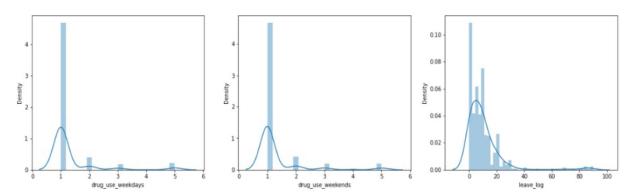


Fig 2: Analysis of Drug usage in Weekdays, Weekends and the Leave Logs

From the above figure, we showed that most of the students do not use the drugs. Few students use the drug weekly one or two times that also makes them to take leave.

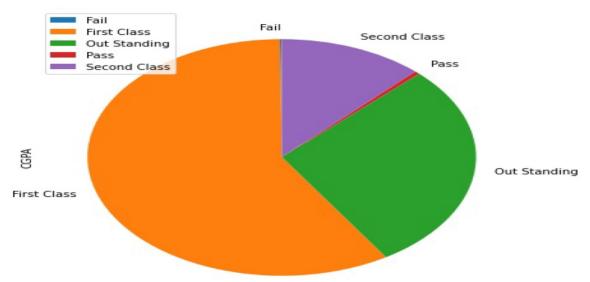


Fig 3: Analysis of the Grades obtained by the students

From the results obtained, it is cleared that the students obtained first class than the other classes of grades.

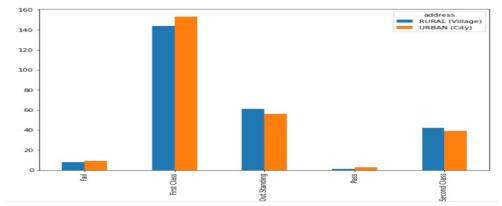


Fig 4: Performance Analysis of Rural and Urban Students

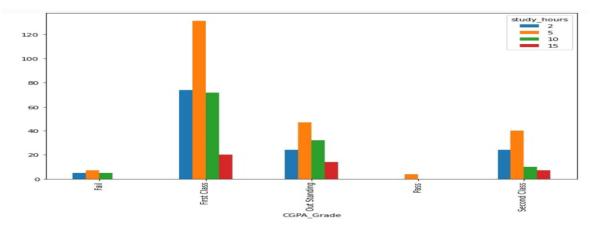


Fig 5: Analysis of Study Hours of the Students

The present study confirms the finding that on an average of five hours of study will make the students to score good grades. Here, the study also reveals that the study hour does not give the improvement in grades. Full focus of study is needed than the hours.

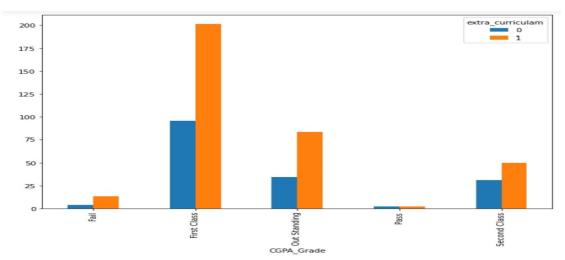


Fig 6: Analysis of the Extra Curricular activities of the students

Another promising finding was that the students with extra curricular activities are performing better than the students without extra curricular activities.

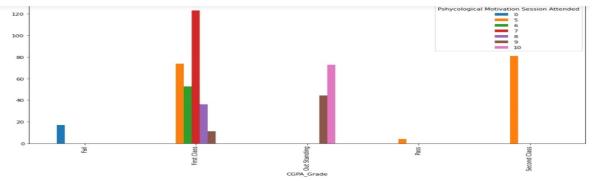


Fig 7: Analysis of Psychological Motivation Sessions attended

It is worth discussing the psychological motivation sessions improve the performance of students in terms of CGPA revealed by the above results.

Table-3 Confusion matrix of Linear Regression model

This suggests that the confusion matrix is a summary of prediction results on a classification problem. Here, the confusion matrix represents counts from predicted and actual values. We have created a confusion matrix for linear regression classifier.

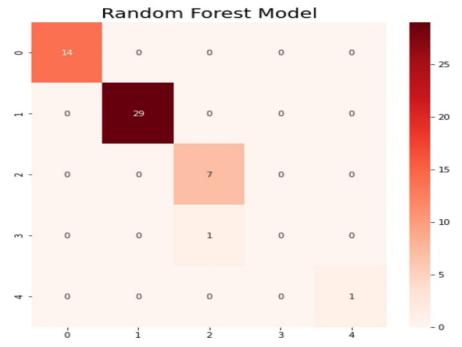


Table 4 Confusion matrix of Random Forest Model

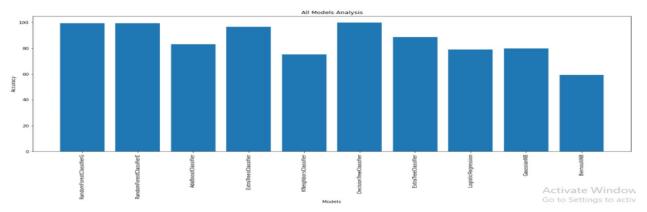


Fig 11: Performance Analysis of different machine learning algorithms

By analyzing the performance in terms of accuracy by different machine learning algorithms, we hope that decision tree classifier performs very well in the prediction of academic performance of college students with bipolar disorder

| Measure s | Logistic Regressio n | Random Forest Classifier Gini | Random Forest Classifier Entropy | Decisi on Tree | KNeighbors Classifier | Ada Boost Classifier | Extra Tree Classifier | Extra Tree Classifier N | GaussianN B | BernoulliN B |
|--------------|----------------------------|--|---|----------------------|--------------------------|----------------------------|-----------------------------|----------------------------------|----------------|-----------------|
| Precision | 0.705341 | 0.963942 | 0.963942 | 1.0 | 0.701735 | 0.725572 | 0.906050 | 0.906050 | 0.931818 | 0.443794 |
| Recall | 0.807692 | 0.980769 | 0.980769 | 1.0 | 0.730769 | 0.846154 | 0.923077 | 0.961538 | 0.920435 | 0.457892 |
| F1Measure | 0.749352 | 0.971795 | 0.971795 | 1.0 | 0.711844 | 0.778555 | 0.912305 | 0.952086 | 0.920435 | 0.457892 |
| Sensitivity | 0.807692 | 0.980769 | 0.980769 | 1.0 | 0.730769 | 0.846154 | 0.923077 | 0.961538 | 0.923077 | 0.538462 |
| Specificity | 0.807692 | 0.980769 | 0.980769 | 1.0 | 0.730769 | 0.846154 | 0.923077 | 0.961538 | 0.923077 | 0.538462 |
| R Squared | 0.616780 | 0.970522 | 0.970522 | 1.0 | 0.410431 | 0.675737 | 0.882086 | 0.941043 | 0.882086 | 0.115646 |
| MAE | 0.211538 | 0.019231 | 0.019231 | 0.0 | 0.307692 | 0.173077 | 0.076923 | 0.038462 | 0.076923 | 0.500000 |
| MSE | 0.250000 | 0.019231 | 0.019231 | 0.0 | 0.384615 | 0.211538 | 0.076923 | 0.038462 | 0.076923 | 0.576923 |
| RSME | 0.500000 | 0.138675 | 0.138675 | 0.0 | 0.620174 | 0.459933 | 0.277350 | 0.196116 | 0.277350 | 0.759555 |
| TPR | 29.00000 | 29.00000 | 29.00000 | 29.0 | 25.000000 | 29.00000 | 29.00000 | 29.00000 | 26.000000 | 25.000000 |
| TNR | 13.00000 | 14.00000 | 14.00000 | 14.0 | 9.000000 | 14.00000 | 13.00000 | 14.00000 | 14.000000 | 2.000000 |
| FPR | 1.000000 | 0.000000 | 0.000000 | 0.0 | 5.000000 | 0.000000 | 1.000000 | 0.000000 | 0.000000 | 11.000000 |
| FNR | 0.000000 | 0.000000 | 0.000000 | 0.0 | 3.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 3.000000 |
| Accuracy | 0.807692 | 0.980769 | 0.980769 | 1.0 | 0.730769 | 0.846154 | 0.923077 | 0.961538 | 0.923077 | 0.538462 |

Table-5 Performance result for Classifiers

Table 5 clearly shows the performance of every classifier based on the precision, recall, F1 measure, sensitivity, specificity, R squared, mean absolute error, mean squared error, root mean squared error, true positive rate, true negative rate, false positive rate, and false negative rate. These measures are useful for comparing the classifiers based on accuracy. From the above table, we found that the Decision tree classifier outperforms in all the aspects than the other classifier within the collected student's dataset. BernoulliNB classifier produces low values

VII. CONCLUSIONS

The paper presented different prediction models to predict the academic performance of bipolar disorder students based on Address, study hours, extra curriculum, internet access, family interaction, drug use weekdays, drug use weeends, leave log, CGPA, Psychological motivation session attended, CGPA Grade. This study is however limited to the college students of Jamal Mohamed College and other few college students. observation, it is found that the performance of a decision tree classifier is best than that of different algorithms applied in this research. The performance of every classifier compared is based on the precision, recall, F1 measure, sensitivity, specificity, R squared, mean absolute error, mean squared error, root mean squared error, true positive rate, true negative rate, false positive rate and false negative rate. This study is helpful for the educational

institutions for the prediction of academic performance of college students affected with bipolar disorder.

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