

Utilization of Machine Learning for Students in Education Sector

Chiranjeevi Kommula¹, Dr. Balusupati Veera Venkata Siva Prasad²

¹Research Scholar, Career Point University

²Research Supervisor, Career Point University

Abstract - The events of 2020 have taught us that culture is already fragile, and that it is vulnerable to events that shift the paradigms that rule it quickly. A pandemic like Coronavirus disease 2019 has shown this; this global emergency has transformed the way citizens connect, chat, learn, and function. The need to extract useful information from data becomes more pressing. In data mining and data analytics, methods and approaches that were once mostly seen in academic labs are now being adopted by forward-thinking businesses to produce market insight and improve decision-making. It's not easy to separate reality from fiction and recognize study opportunities and realistic implementations as analytics and data mining projects in education become more common. Learning analytics (LA) as a field stays in its earliest stages. Large numbers of the strategies now unmistakable from professionals have been drawn from different fields, including HCI, computer science, statistics, and learning sciences. Machine learning and data analytics are proposed methods that can help remove data and discover important examples inside the gathered data. In this work, the field of e-learning is researched regarding definitions and attributes. This Article dissects the Utilization of Machine Learning for Students in Education Sector.

Index Terms- Educational Data Mining, artificial intelligence; machine learning; data analysis, Learning analytics, etc.

I. INTRODUCTION

Currently, society is dealing with a wellness crisis that has altered how people work. The Coronavirus Outbreak of 2019 (COVID-19) has exposed the vulnerability of all sectors, including wellness, education, and industry. There has been an impact on every aspect of society; moreover, it is the responsibility of universities and their academic teams to address many of these flaws and develop comprehensive frameworks focused on what has been

learnt from this crisis. It is important to consider the resources that have helped us to fight this epidemic and that have acted as a conduit to maintain those places open and usable, which are required for society's growth and survival. These techniques are information and communication technology (ICT), which have made it possible to conduct most operations remotely and safely. It's worth noting that what occurred has altered our perspective on life.

In the 21st century, AI has become a significant space of examination in all fields: science, education, Engineering, business, medicine, accounting, marketing, finance, stock market, economics, and law, among others. The scope of AI has developed tremendously since the intelligence of machines with machine learning abilities significantly affects business, governments, and society. The objective that this record might want to accept that isn't the trigger of an abrupt multiplication of a generally united area, however it is trusted that this exploration could be a significant scholarly apparatus for both the pulling together of the work and setting out new learned open doors.

E-learning is one of the later fields that is adding to the lot of produced data. E-learning can be characterized to be the utilization of electronic devices and technology for learning new information and skills. The multiplication of technology all through the world and the blast in information access has made distance learning more mainstream lately. Distance learning is one segment of the e-learning measure as it permits individuals to share information in spite of geographical limits and impediments. Social network investigation, recommender systems, just as student achievement and prescient models depend vigorously on data drawn from these systems. Learning analytics projects over the previous decade have depended on sensibly very much organized data drawn from learning

management systems (LMS) and student information systems (SIS).

The developing variety of educational data manages the cost of scientists a chance to investigate new methodologies, effectively practically speaking in fields, for example, artificial intelligence and machine learning, to break down a more extensive range of student data than is as of now being thought of. In this work, the combination of Machine Learning, data Analytics, and LA is dissected to improve an online education demonstrate and along these lines improve student learning.

II. MACHINE LEARNING

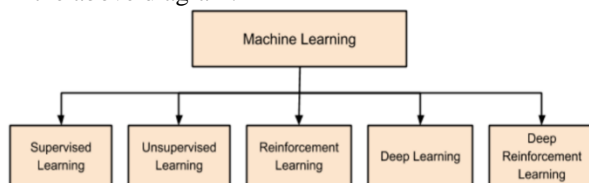
Artificial Intelligence (AI) has already outpaced the hysteria of blockchain and quantum computing in recent years. This is attributed to the reality that the average person has access to vast computer facilities. Developers are also using this information to create new Machine Learning models and to retrain current models for improved efficiency and outcomes.

Machine learning comes in a variety of forms. Machine Learning has been one of the mainstays of computer technology over the last two decades, and with it, a very important, but mostly secret, part of our lives. With ever-increasing volumes of data being accessible, there's cause to think that smart data processing would become much more prevalent as a critical component of technical advancement.

Uses of Machine Learning

- Credit card fraud detection
- Spam filtering
- Detecting faces in images
- Digit recognition on checks, zip codes
- Recommendation system
- MRI image analysis
- Scene classification
- Handwriting recognition
- Search engines

Machine learning evolved from left to right as shown in the above diagram:



III. LEARNING ANALYTICS

Learning, like knowledge, is impossible to describe simply since it encompasses so many different mechanisms. 'To acquire insight, or comprehension of, or ability in, through observation, training, or experience,' according to a dictionary description, as well as 'modification of a behavioral pattern through experience.' Animal and human learning were studied by zoologists and psychologists. Animal and computer learning have a number of similarities. Many machine learning approaches are undoubtedly derived from psychologists' attempts to use statistical methods to refine their theories of animal and human development. It's also possible that the principles and approaches being discussed by machine learning researchers will shed light on certain facets of biological learning.

In terms of computers, we might assume that they learn if their configuration, program, or data is changed in such a way that their predicted future output increases (based on their inputs or in reaction to external information). Machine learning is one of the hottest fields of computer technology, with a wide range of implementations.

The assessment, processing, study, and reporting of data regarding learners and their contexts for the purposes of understanding and maximizing learning and the situations in which it happens is referred to as LA. At the moment, educational data mining is primarily concerned with the creation of new methods for detecting trends in data. These habits are all for studying microconcepts such as one-digit multiplication, subtraction with carries, and so on. Learning analytics relies on incorporating methods and strategies to broader rates, such as in classes and at colleges and postsecondary universities, as opposed to data mining, at least as it is currently compared with data mining.

IV. EDUCATION AND MACHINE LEARNING

There are plenty of other business implementations of machine learning, lot of them are in education area. Some of interesting areas are:

Fairly Test and Grade Students (Machine learning may assist in the development of computerized adaptive assessments.) The machine learning-based evaluation

provides teachers and educators with continuous guidance about how the student performs, the help they need, and their success against their learning objectives.)

Predict Student Success (Predicting student performance is a fantastic application of machine learning. The machine learning model will identify shortcomings in each pupil by 'learning' about them and suggesting ways to better, such as adding further lectures or reading more literature.)

Things to support teachers and institutions (Machine learning based algorithms can help with classification of students handwritten assessment papers)

Boost Retention (Machine learning, such as learning analytics, may also aid retention.) By finding students who are 'at risk,' schools will reach out to them and provide them with the support they need to succeed.)

4.1 Analytics in education

'No one method of teaching fits well for all students,' educators and scholars have discovered. Since students' motivations, desires, and skills vary, utilizing a diverse range of methods is critical.' Teachers and school leaders in several countries already have access to growing amounts of data and data resources, such as personalized report dashboards. By offering a consistent picture of student success, these data and reports have had a positive impact on teaching and learning. However, data tools do not often have advice about a student's next moves in learning, and for best results, they take advanced expertise as well as time. When it comes to training different students for a changing world, schools and academic programs pose challenges. Collecting data on student learning and system success is a part of tracking and measuring their development. The resources and capacity to use data to optimize results have not always kept up with the growth and complexity of these datasets.

Schools and educators also have access to innovative data and analytics resources that will assist them in dealing with the complexities of schooling. Educators can better understand students' static data (for example, demographics; previous achievement) and dynamic data (for example, pattern of online logins; quantity of conversation posts) using powerful data visualization software that can be personalized for each person, address unique questions, and automatically update to include current knowledge. Today's sophisticated analytics technologies, such as

deep learning, will display students' progress (for example, at risk; high achiever; social learner), allowing for more timely measures (for example, offering extra social and academic support; presenting more challenging tasks).

The advantages of data visualization, analysis, and prediction benefit:

- Parents get detailed reports on student progress
- Students get feedback on their pattern of performance in learning systems, and who may be assessed more frequently in ways that better guide progress and give them a personalized learning experience
- Content designers and curriculum managers get data on content usage and relationships between content and learning
- Teachers get detailed reports on all students, as well as relative effectiveness of lessons and content, freeing them from low-order assessments to focus on more complex feedback
- School system leaders get data across campuses year to year
- School leaders get student progress data, teacher effectiveness data, and school-level outcomes. Facility factors like busing, buildings, schedules, and activities can be factored into learning. Staff factors like professional learning and credentials can be analyzed
- Education researchers who will get ongoing insights into impacts of practices and conditions at large scale
- Policymakers get outcomes associated with different school and community conditions
- As education analytics advances, a wider range of data can be integrated for more refined and personal guidance.

V. UTILIZATION OF MACHINE LEARNING FOR STUDENTS IN EDUCATION SECTOR

Four categories that clearly discuss the use of ML in Education sector:

Improving student retention: Like we said before, schools will spot and contact the kids early by finding students at risk and helping them succeed. The retention of students is an important feature of many registration schemes. Almost all segments of academic

or school measurements are affected: credibility, finance, rating. In particular, the retention of students has been one of the most critical aspects for managers in universities. There are few research that have established models to forecast and explain the reasons behind the declining number of students

Grading students: Machine learning can graduate students by eliminating human prejudices. Some recent examples include the usage of supervised machine learning in text classification for the prediction of final courses for students in certain classes, and the ability to recognise students at risk of course failure with classified messages from ML. In addition, it is intended to enhance the evaluation of educational problem solving by using language technology and computer statistic training approaches to automatically evaluate the natural language answers of pupils. A great illustration of how students learn to use machines by matching their behavior with a model of expert behaviour.

Testing students: The machine learning evaluation gives instructors, students and parents continuous updates on how the student succeeds, how they require guidance and how they advance towards their learning objectives. In a thesis, a teaching method for students to construct the right evidence in propositional or predicate logic. In addition to traditional technologies, they used animations focused on specifically chosen demonstrators and step-by-stage solutions, including slides-supported demonstrations and drills. To assess the awareness of students, a questionnaire was produced that captured the entire phase of a logical proof creation. A student generated evidence and addressed questions from the questionnaire. They outlined and explained the concept of the questionnaire. In addition to supervised machine learning algorithms, they used frequent subgraph mining to automatically evaluate the accuracy of the tests.

Predicting student performance: Machine learning (regarding the amount of studies in science databases) probably benefits greatly from its capacity to anticipate student success. The technology will detect shortcomings and recommend strategies to strengthen them through “learning” about each pupil, such as extra practice testing. This seems to be a very hot topic in research; there have been several studies in this field in recent years, as we said before. For example, an

analysis uses the machine learning method called the recursive clustering strategy to divide the students in the curriculum into groups based upon their success in the prerequisite classes, their core requirement and their current course results. Students present in the lower classes are taken into account as they are very susceptible to failure. In another important research in this category, the writers suggested a novel model to classify students into three groups to assess their academic skills and allow them to develop their techniques of learning.

VI. CONCLUSION

Researchers will use emerging technologies, which are now in use in fields like artificial intelligence and deep learning, to study a wider range of learner data than is typically being regarded, thanks to the increasing diversity in educational data. Both the importance of data as a platform for science discovery and the advantages of statistical models and hypotheses for educational technologies and study have been noted by educational researchers. Educational scholars have been able to produce data-driven hypotheses regarding learning and learning results thanks to recent developments in statistical approaches such as machine learning. Educational data mining and learning analytics are being utilized to conduct analysis and develop models in a variety of fields that have the potential to impact online learning structures. Educational data mining and learning analytics have the ability to render previously unknown, overlooked, and therefore unactionable data available. The research was conducted using many applicable sources, although not all of them were used, so this can be called a study restriction. Furthermore, it is possible that any of the related experiments would be overlooked by accident.

REFERENCES

- [1] Alam, M. M., Mohiuddin, K., Das, A. K., Islam, M. K., Kaonain, M. S., & Ali, M. H. (2018, March). A Reduced feature based neural network approach to classify the category of students. In Proceedings of the 2nd International Conference on Innovation in Artificial Intelligence (pp. 28-32). ACM.

- [2] Chai, K. E., & Gibson, D. (2015). Predicting the Risk of Attrition for Undergraduate Students with Time Based Modelling. International Association for Development of the Information Society.
- [3] Chen, C. Personalized E-learning system with self-regulated learning assisted mechanisms for promoting learning performance. *Expert Syst. Appl.* 2009, 36, 8816–8829.
- [4] Ciolacu, M., Tehrani, A. F., Beer, R., & Popp, H. (2017, October). Education 4.0—Fostering student's performance with machine learning methods. In *Design and Technology in Electronic Packaging (SIITME), 2017 IEEE 23rd International Symposium for* (pp. 438-443). IEEE.
- [5] Colin, H.; Donnelly, I.A. Ambient intelligence: Technologies, applications, and opportunities. *Proc. Annu. Meet. ISSS 2017*, 91, 399–404.
- [6] Đambić, G., Krajcar, M. & Bele, D. (2016). Machine learning model for early detection of higher education students that need additional attention in introductory programming courses. *International Journal of Digital Technology & Economy*, 1 (1), 1-11
- [7] Jia, J. W., & Mareboyana, M. (2014). Predictive models for undergraduate student retention using machine learning algorithms. In *Transactions on Engineering Technologies* (pp. 315-329). Springer, Dordrecht.
- [8] Lee, S.J.; Lee, H.; Kim, T.T. A study on the instructor role in dealing with mixed contents: How it affects learner satisfaction and retention in e-learning. *Sustainability* 2018, 10, 850.
- [9] Luckin, R., Holmes, W., Griffiths, M., & Forcier, L. B. (2016). *Intelligence unleashed: An argument for AI in education.*
- [10] Popoola, S.I.; Atayero, A.A.; Badejo, J.A.; John, T.M.; Odukoya, J.A.; Omole, D.O. Learning analytics for smart campus: Data on academic performances of engineering undergraduates in Nigerian private university. *Data BR* 2018, 17, 76–94.
- [11] Syeda Farha Shazmeen, S.F.S. Performance Evaluation of Different Data Mining Classification Algorithm and Predictive Analysis. *J. Comput. Eng.* 2013, 10, 1–6.
- [12] Villegas-Ch, W.; Luján-Mora, S. Systematic Review of Evidence on Data Mining Applied to LMS Platforms for Improving E-Learning. In *Proceedings of the International Technology, Education and Development Conference, Valencia, Spain, 6–8 March 2017*; pp. 6537–6545.
- [13] Villegas-Ch, W.; Palacios-Pacheco, X.; Buenaño-Fernandez, D.; Luján-Mora, S. Comprehensive learning system based on the analysis of data and the recommendation of activities in a distance education environment. *Int. J. Eng. Educ.* 2019, 35, 1316–1325.
- [14] Villegas-Ch, W.; Palacios-Pacheco, X.; Luján-Mora, S. Artificial intelligence as a support technique for university learning. In *Proceedings of the IEEE World Conference on Engineering Education (EDUNINE), Lima, Peru, 26 May 2019*; pp. 1–6.
- [15] Wu, J. Y., Hsiao, Y. C., & Nian, M. W. (2018). Using supervised machine learning on large-scale online forums to classify course-related Facebook messages in predicting learning achievement within the personal learning environment. *Interactive Learning Environments*, 1-16