

## Using Machine Learning in an LMS to Implement the Online Education Model

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**Abstract:** The demand for Learning Management Systems (LMS) that provide scalable, adaptable, and personalized learning experiences has increased due to the explosive expansion of online education. While developing Machine Learning (ML) techniques allow for automated decision-making, tailored learning paths, predictive analytics, and intelligent evaluation mechanisms, traditional LMS platforms mostly concentrate on content delivery and administrative tasks. In order to promote a strong online education model, this study investigates how machine learning (ML) can be included into an LMS. System architecture, machine learning applications, implementation difficulties, and potential research avenues are presented. The results highlight how ML-driven LMS solutions can revolutionize learning outcomes, engagement, and institutional effectiveness.

**Keywords:** *Analysis of Data; Artificial Intelligence; Machine Learning; Online Education; Education Content*

### INTRODUCTION

A vital part of contemporary learning ecosystems is now online education. Because they facilitate communication, content management, student evaluation, and instructional delivery, LMS systems are essential to this shift. However, student data is rarely completely utilized by standard LMS platforms for adaptive or personalized learning. The intelligence and processing capacity needed to create more dynamic, responsive, and successful online learning environments are provided by machine learning.

LMS platforms that incorporate machine learning (ML) can increase student engagement, forecast performance, automate administrative duties, and give teachers real-time decision support. This study looks at how the online education paradigm can benefit from ML-based improvements.

The world's current normalcy brings with it new challenges that must be overcome and that all regions must embrace through the use of ICT and new developments. One of the sectors of society that

has integrated these advances into its operations for a long time is advanced education. Despite the significant drawbacks, this combination has allowed education to continue. Nevertheless, it is critical to recognize the problems that have arisen and adopt flexible educational models that integrate fresh and improved innovations that enable students to continue their education under any conditions. Returning to particular concepts and tools that have been disregarded is crucial to achieving this aim.

### PRIOR WORKS

Data evaluation runs the risk of looking at a lot of data to make decisions about which information to choose or to expand on a particular topic. To provide precise closures that aid in reaching the suggested goals, data evaluation runs the data through a number of errands.

Similar to how data evaluation is utilized in the sciences to support or refute preexisting models or hypotheses, it is employed in a variety of endeavors to empower associations and relationships to pursue superior business selections. Its augmentation, inspiration, and examination-related considerations characterize the distinction with data extraction. Data extractors use intricate programming to display enormous educational information in order to

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identify hidden models and disseminate hidden correlations. Deriving is the most well-known method for determining a surmising subordinate based solely on the researcher's knowledge. The following districts typically employ data examination:

- Marketing: Data analysis has essentially been utilized to anticipate and organize buyer leads.
- HR: Data analysis is also very helpful for organizations to maintain a positive work atmosphere and see expected delegates.
- Scholastics: Data assessment is often used in tutoring to choose new students and evaluate their performance.

## LITERATURE REVIEW

### • Learning Management Systems

An LMS typically supports content delivery, learner tracking, assessments, and administrative tasks. Popular LMS platforms include Moodle, Blackboard, and Canvas. These systems rely heavily on predefined rules and lack adaptive intelligence.

### • Machine Learning in Education

Machine Learning applications in education include classification models for predicting outcomes, recommender

systems for personalized content, NLP for automated grading, and clustering algorithms for identifying learning patterns.

### • Online Education Models

The online education model emphasizes flexibility, accessibility, learner autonomy, and data-driven instructional methods. Integrating ML can strengthen these principles through personalization and real-time analytics.

## MODEL OF ONLINE EDUCATION

The development of ICT has created countless opportunities for educational activities that allow everyone to access high-quality education regardless of their location or time. Time and distance have undoubtedly been eliminated as a barrier to education and learning by the numerous options that people now enjoy. Online training is a distance learning approach that uses mechanical equipment to provide an educational experience in a computerized environment called a virtual study hall that may be accessed via a Web association. It has the advantage of being a no concurrent focus on the

model, where the days and hours of the week are arranged for cooperation with the teacher. Because of the hectic pace of daily life, online education has emerged. Habit that society currently goes through [29]. For the geological, professional, or familial place of some people, online learning achieves a common educational objective, without the boundaries of space or time.

## THE ENVIRONMENT'S IDENTIFICATION

A Madhya Pradesh school that provides two survey modalities took part in this study. Since learning depends on the teacher's attitude and experience, the primary approach meets the requirements of a traditional methodology in that it is intimate and personal.

The learner becomes an eyewitness to her own learning and should agree with recently decided plans. Furthermore, the instructor becomes the deciding element for what they need to know and how to do the assignment. Consequently, the teacher has more control over determining each student's performance. This evident evidence thus favors the educator's rules unilaterally, which is not typical of an ideal learning model. The foundation of the work is the mechanical design of this interaction, which has enabled the integration of data innovation (IT). This web-based guidance model's process is designed for individuals who are unable to follow an eye-to-eye method due to their schedules and obligations. This methodology creates degree-based institutions and courses by starting with a learning management system (LMS).

The student should adhere to three mandatory exercises in the online course: task improvement, evaluations, and conversational cooperation. In order to enhance these activities, the student will complete a module in which he will locate all of the materials deemed important by the course's instructor and designer.

Furthermore, the instructor becomes the deciding element for what they need to know and how to do the assignment. Consequently, the teacher has more control over determining each student's performance. This evident evidence thus favors the educator's rules unilaterally, which is not typical of an ideal learning model. Online guidance is the second model that the school provides. This model has been developed and refined over the last ten years.

Furthermore, its enhancement is as close as possible to the virtual guidance model. The foundation of the work is the mechanical design of this interaction, which has enabled the integration of data innovation (IT). This web-based guidance model's process is designed for individuals who are unable to follow an eye-to-eye method due to their schedules and obligations. This methodology creates degree-based institutions and courses by starting with a learning management system (LMS). The student should adhere to three mandatory exercises in the online course: task improvement, evaluations, and conversational cooperation. In order to enhance these activities, the student will complete a module in which he will locate all of the materials deemed important by the course's instructor and designer.

The school's employment of a mechanical structure for this process becomes advantageous to make this work better. It has been set up to support a large number of administrations and trades. The IT sector handles issues including information security and care. This ensures that the effort would focus on merging LMS, information inquiry, and computer-based intelligence without sacrificing data. The IT design that is shown in Figure 1 brings the school together. There are layers in this plan, and one more has been incorporated. That are in charge of evaluating the data.



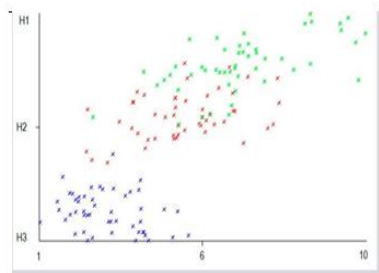
**Figure 1: Technological architecture of online education modality**

## DISCUSSION AND RESULTS

The new normal that people live in forces foundations to look for new models that meet the needs of people. This idea is taken into consideration, and the goal of this paper is to further develop a model for online instruction. The joining of advancements transforms into the early phase to additionally foster guidance and screen student execution. It is important to keep in mind that the current reality has made online, virtual, or hybrid learning models the usual way to continue with higher education. This work is applied on the designing and underpinning of the school that participated in the survey. Since sending the majority of the framework enables the grouping of

tasks on the AI model's plan, this is seen as a benefit. If there is a need to change any layer of the plan, it is basically revived without the need to make higher particular, human, or financial costs. The observing of student performance is improved when these advancements are combined. The majority of the time, this largely depends on the models of the educator or those in charge of learning. With this model, the noticing doesn't have human performers, the structures are liable for finishing an interminable examination of each and every student, and the computer based intelligence model will attempt to perceive the cases that have the most raised risk of low educational execution. This component grants you age of an early notification ahead of time that is by and by settled when the educational checking division knows a particular number of grades. Early ID of the total model allows the period of projections considering the student's arrangement of encounters. For instance, the framework considers students with difficulties in the subject of prologue to analytics as potential cases with difficulties in math I and subjects whose fundamental is prologue to analytics. This examination can be astoundingly shallow; in any case, the system could conclude a possible example of emphasis by taking apart the focuses that make up a subject.

For the idea of activities, simulated intelligence is familiar with the student's presentation in each activity. The decision is then made based on the best outcomes the student achieves in each action. For instance, cases have been identified in which type exercises, which are quick assessments based on real and fake things, do not meet the needs of a particular group of understudies. The model identifies these groups and suggests various exercises to the course creator. For this, the headway of dynamic learning is taken as a substance. A wide range of exercises that AI suggests to the student based on their needs are fostered in this type of learning.



**Figure 2: Data analysis of the activities developed in an online education model**

The result obtained by big data is taken by the AI to feed machine learning and learn about this data for decision-making. The AI model integrated the analysis, the data from the LMS in relation to the time of dedication of the students to the reading of the teacher's resource, and the data from a survey carried out on the students, where the time they had to answer each question was discussed. The data from this analysis was subjected to the naive Bayes data mining algorithm with the results presented in Table 1.

**Table: 1 Stratified Cross- Validation**

Correctly Classified Instances	48	94.12%
Incorrectly Classified Instances	3	5.88%
Kappa statistic	0.9113	
Mean absolute error	0.0447	
Root mean squared error	0.1722	
Relative absolute error	10.04%	
Root relative squared error	36.42%	
Total Number of Instances	51	

The algorithm performed the analysis of 51 instances to identify the reason why the scores in the evaluations present a performance below the expected. Of the 51 instances, 48 were classified as correct, with 94.1176%.

The figure demonstrates that the machine learning model's data analysis and learning significantly improved the evaluations. At the point when vital, the model adds a more prominent number of factors to the examination and pursues a choice thinking about the outcomes. In this exercise, data from the LMS and a student survey were identified. The model permits change of the loads to ensure compelling navigation. Additionally, effective

corrections can be made prior to an event becoming a problem thanks to the model's quick action. This thought is made when assessments were doled out in a similar period.

## CONCLUSION

In online education, machine learning has revolutionary possibilities for improving LMS functionality. ML-driven LMS solutions facilitate enhanced learning outcomes, engagement, and institutional efficiency by facilitating automation, personalization, predictive analytics, and intelligent evaluation. Even though there are obstacles, incorporating ML into LMS architecture is a big step in creating educational ecosystems that are prepared for the future.

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