

Research on Student Learning Behavior Analysis and Precision Teaching Driven by Big Data

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Abstract: *With the rapid development of big data technology, the education sector has also encountered unprecedented opportunities. Traditional teaching models are gradually shifting toward data-driven personalized instruction, allowing teachers and educational administrators to leverage vast amounts of learning data to gain deep insights into students' learning behaviors, optimize teaching strategies, and implement precise teaching. This paper aims to investigate student learning behavior analysis and precise teaching driven by big data. Through comprehensive analysis of student learning trajectories, online participation, classroom interactions, and assignment data, this study constructs various learning behavior analysis models, revealing students' learning habits and personalized needs, thereby providing teachers with robust evidence for instructional interventions. Additionally, based on these analyses, this paper proposes precise teaching strategies, including personalized learning path planning, dynamic resource recommendations, and real-time feedback mechanisms, further enhancing teaching efficiency and learning outcomes. In the future, with the integrated development of artificial intelligence technology, precise teaching is expected to achieve a high level of automation and intelligence; however, data privacy and ethical issues must also be given adequate attention.*

Keywords: *Big Data; Student Learning Behavior Analysis; Precise Teaching; Personalized Learning; Artificial Intelligence; Data Privacy*

Introduction

With the rapid advancement of information technology, particularly the increasing maturity of big data and artificial intelligence technologies, the education sector is undergoing profound transformations. Traditional education primarily relied on experience and intuition, whereas, in the era of big data, educational decision-making and management increasingly depend on data analysis. Big data technology, through real-time collection and processing of students' learning data, provides educators with unprecedented deep insights, making personalized learning feasible. Student learning behavior analysis plays a crucial role in this process, as it not only helps teachers accurately identify students' learning progress and difficulties but also provides data support for precise teaching. Therefore, studying big data-driven student learning behavior analysis and precise teaching not only aids in optimizing educational resource allocation and improving teaching effectiveness but also offers new directions for enhancing educational equity and quality. Against this backdrop, this paper explores the analysis of student learning behavior and data-driven precise teaching, aiming to provide practical theoretical support and guidance for the education sector.

1. Deep Integration of Big Data and Education

1.1 Educational Context in the Era of Big Data

The rapid development of big data technology has brought unprecedented opportunities and challenges to global educational systems. In the era of big data, education is gradually transitioning from traditional experience-driven approaches to data-driven methods. By analyzing vast amounts of learning data, educators can gain a deep understanding of students' learning behaviors and utilize data analysis to optimize teaching models, thereby promoting the realization of personalized education. In this process, the use of online learning platforms, classroom interactions, assignment submissions, and other elements generates significant data, which becomes a critical resource in modern education. For example, data such as students' learning trajectories, online participation rates, assignment submission

frequencies, and test scores provide teachers with a comprehensive view of student learning performance. By integrating this information, teachers can more accurately grasp students' learning progress and identify bottlenecks, allowing for adjustments in teaching strategies^[1].

In this context, big data technology not only transforms classroom teaching but also changes the ways in which educational management and decision-making occur. Educational administrators and policymakers can make scientifically informed decisions based on big data, shifting educational management from static, experience-based judgments to dynamic, precise optimizations. For instance, schools can analyze course registration and performance data to flexibly adjust course offerings in response to students' needs and ability differences. Simultaneously, based on real-time data regarding student learning performance, teachers can provide timely interventions, offering personalized tutoring and learning recommendations. This data-driven educational transformation significantly enhances teaching efficiency, quality, and student learning experiences.

1.2 Current Research Status on Student Learning Behavior Analysis

Student learning behavior analysis, as an important application area of big data in education, has garnered widespread attention in recent years. Current research mainly focuses on how to effectively collect, process, and analyze multi-source data generated during the learning process. This data includes, but is not limited to, data from online learning platforms, records of classroom behaviors, and test scores. By analyzing this data, researchers aim to understand students' learning behavior patterns, reveal their learning habits, and predict academic performance and potential issues. To achieve this goal, modern learning behavior analysis typically relies on cutting-edge technologies such as machine learning, data mining, and statistical analysis. These tools can handle complex, heterogeneous learning data and reveal students' learning characteristics through techniques such as classification, clustering, and association analysis, thus helping educators implement targeted instructional interventions.

Despite the significant progress in learning behavior analysis, numerous challenges remain. First, integrating and processing multi-source data poses a challenge in current research, particularly with heterogeneous data from different platforms and formats. Achieving effective integration at the technical level requires further exploration. Second, addressing individual differences is also a crucial issue. Factors such as students' learning styles, cognitive abilities, and emotional states vary widely, necessitating learning behavior analysis models with greater adaptability to individual needs. Furthermore, data privacy concerns are especially sensitive in the education sector. Balancing the effective use of learning data for analysis with the need to protect student privacy presents both ethical and technical challenges for researchers. Future studies in learning behavior analysis will increasingly emphasize the precision of personalization and the scalability of models, along with deeper exploration of data ethics and privacy protection^[2].

1.3 Theoretical Foundations of Precise Teaching

Precise teaching represents one of the core applications of big data technology in education, emphasizing the implementation of personalized teaching strategies through data-driven learning feedback and behavior analysis. Its theoretical foundations primarily derive from personalized learning theory and constructivist learning theory. Personalized learning theory advocates for recognizing that each student has unique learning characteristics, interests, and needs, and that education should be designed to accommodate these individual differences with flexible and diverse teaching plans. Constructivist learning theory emphasizes that learning is an active process of knowledge construction, with the teacher's role being more of a guide and supporter. Through interaction and feedback with students, teachers help them build their own knowledge systems. Big data technology provides robust support for the practical application of these two theories. By analyzing students' learning behaviors and knowledge mastery, precise teaching can adjust teaching content, pace, and methods based on each student's characteristics.

Data-driven precise teaching is not only theoretically rooted in personalized learning and constructivism but also concretely implemented through data technologies in practice. By utilizing big data analysis, teachers can monitor students' learning progress and behaviors in real time, promptly identifying issues and generating personalized learning feedback. This feedback includes information on students' mastery of specific knowledge points, changes in learning habits, and levels of engagement. Based on this feedback, teachers can implement targeted instructional interventions, adjusting teaching content and methods to better meet students' learning needs. In the future, with ongoing advancements

in data analysis and artificial intelligence technologies, the degree of personalization in precise teaching is expected to increase, enhancing not only the flexibility of teaching but also significantly improving learning effectiveness and efficiency.

Moreover, precise teaching is not limited to traditional classroom settings; it also permeates online education and blended learning environments. By leveraging big data, precise teaching can facilitate cross-platform applications across various learning scenarios, making instruction more flexible and intelligent. Utilizing learning analytics tools, teachers can continuously monitor students' learning dynamics and adjust teaching strategies in real time, ensuring that every student receives optimal learning experiences tailored to their needs and abilities. In this teaching model, educators can not only assist students in resolving specific learning issues but also predict future learning needs through data analysis, enabling the development of long-term teaching plans^[3].

2. Analysis of Student Learning Behaviors Based on Big Data

2.1 Sources and Types of Learning Behavior Data

In a big data-driven education system, sources of student learning behavior data are extensive and diverse, encompassing data from online learning platforms, classroom teaching, and individual student academic development. By integrating these data sources, a comprehensive assessment and analysis of the learning process can be conducted, providing a foundation for precise teaching.

Data from online learning platforms represents a crucial source of big data analysis. This data includes students' login times, course browsing history, video viewing durations, assignment completion statuses, and participation frequencies in discussions. Such data can detail students' behavioral trajectories in digital learning environments, reflecting their learning pace, resource utilization, and depth of interaction with course content. Compared to traditional classroom performance, online learning data can more intricately reveal students' autonomous learning habits and preferences, offering a basis for instructional intervention.

Classroom teaching data is derived from teacher-led instructional activities. Modern classrooms increasingly integrate smart devices and teaching tools that can record students' attendance, participation in discussions, and feedback. This data provides important evidence for analyzing students' engagement and learning states in the classroom. Additionally, students' personal academic development data, such as exam scores, semester evaluations, and learning portfolios, can uncover long-term academic performance trends, supporting the assessment of their academic progress and the formulation of personalized learning plans.

2.2 Construction of Learning Behavior Analysis Models

To analyze student learning behaviors in depth, big data analysis relies on various learning behavior analysis models, which help to reveal underlying patterns in the learning process and identify students' personalized needs. One of the core models is the learning trajectory analysis model. This model tracks students' behavioral paths throughout the learning process, analyzing their learning progress, frequency, and knowledge mastery. By analyzing trajectory data, teachers can identify points of stagnation or knowledge weaknesses in students' learning, allowing for targeted instructional support.

The learning interest and engagement model is primarily used to analyze students' interest changes and levels of investment during the learning process. By analyzing data from learning interactions, assignment submissions, and classroom participation, this model can assess the intensity of students' interest in specific content. It can assist teachers in understanding where students demonstrate strong interest in particular knowledge points, enabling the provision of personalized teaching resources and support strategies to enhance learning outcomes.

The learning habits and preferences model captures students' learning habits and resource usage preferences through long-term behavioral data accumulation. By analyzing students' repetitive behaviors and preference selections during learning, teachers can better recommend learning resources that align with students' learning styles, promoting the application of personalized learning modes. This model plays a significant role in planning individualized learning pathways, contributing to improved learning efficiency and experiences.

2.3 Mining and Classification of Student Learning Behavior Features

Feature mining is a crucial step in identifying learning patterns and individual differences during the analysis of student learning behaviors. Time series analysis of learning behavior features is one of the key methods employed. By analyzing students' learning behaviors at different time points, educators can uncover periodic characteristics and stage changes in students' learning behaviors. Time series analysis can help identify peaks and troughs in learning, providing teachers with optimal intervention timing to effectively enhance students' performance.

Classifying the behavior patterns of students with different learning types is another important application of big data analysis. By aggregating and comparing learning behavior data from different students, they can be categorized into types such as efficient learners, task-oriented learners, and passive learners. Behavior pattern classification aids teachers in developing differentiated teaching strategies tailored to various student types, making instruction more personalized and precise^[4].

Clustering analysis techniques hold significant application value in learning behavior data mining. Through clustering analysis, common behaviors among student groups can be classified, forming typical patterns of behavior characteristics. Additionally, clustering results provide a foundation for personalized modeling; by identifying similarities within student groups, teachers can devise more targeted learning plans and interventions, enhancing instructional effectiveness and achieving the goals of precise teaching.

3. Big Data-Driven Precision Teaching Strategies

3.1 Teaching Optimization Based on Learning Behavior Analysis

Big data technology has brought profound changes to the education sector, particularly in the optimization of personalized teaching, where its role cannot be overlooked. By analyzing student learning behaviors, teachers can gain a more precise understanding of each student's learning characteristics and needs, enabling the implementation of more personalized teaching strategies. Personalized learning pathway planning is one of the core components of precision teaching. Based on the analysis of students' learning trajectories, teachers can design personalized learning paths tailored to students' cognitive levels and learning paces. Big data systems can record and analyze students' performance in real time, dynamically adjusting learning paths to ensure that students master knowledge step by step. This dynamic pathway planning helps avoid cognitive overload from excessive learning while assisting struggling students in keeping pace with their studies.

Additionally, dynamic resource recommendations represent an important innovation in big data precision teaching. The system can intelligently recommend learning resources, such as video explanations, exercises, and case analyses, that closely align with students' needs by mining their learning interests, behavioral preferences, and knowledge mastery. Through such dynamic recommendations, students can receive more targeted support for their learning resources, thereby enhancing learning efficiency and maximizing the goals of personalized learning. Moreover, the real-time adjustment of learning progress and goals based on big data can flexibly respond to students' learning performances, dynamically adjusting their tasks and short-term objectives based on real-time feedback. This feature allows both teachers and students to accurately grasp the progress of learning, effectively reducing learning difficulties or inefficiencies caused by mismatched learning paces^[5].

3.2 Teaching Feedback and Intervention Mechanisms

Big data-driven precision teaching relies heavily on efficient teaching feedback and intervention mechanisms. In traditional teaching, feedback often has a delayed effect and is based on experiential judgment, whereas the application of big data makes feedback more timely, precise, and data-driven. A data-based feedback system can perform multi-dimensional analyses of students' learning data, generating detailed feedback reports. These reports provide evaluations of students' learning outcomes, delving into the intricacies of the learning process, including analyses of learning behaviors, knowledge mastery, and changes in study habits. Such refined feedback can help teachers accurately identify each student's learning blind spots and deficiencies, allowing for personalized teaching adjustments. Simultaneously, students can utilize data feedback to understand their learning statuses, adjust their strategies, and enhance their autonomy in learning.

The real-time intervention mechanisms within precision teaching can promptly engage when students encounter learning difficulties, thanks to big data's real-time monitoring capabilities. For instance, when students repeatedly make errors on a particular knowledge point or show significant delays in learning progress, the system can immediately push supplementary resources or suggest additional tutoring. This real-time intervention not only improves the timeliness of teaching but also enhances its personalization, effectively preventing the accumulation of issues that could negatively affect learning outcomes. Optimizing teacher-student interactions based on learning data is also a crucial strategy in precision teaching. Through real-time data analysis, teachers can better understand the specific difficulties and needs students face during the learning process, allowing for more effective interactions. Data-driven teacher-student interactions enhance communication efficiency, enabling teachers to guide students more effectively and improve learning outcomes^[6].

3.3 Future Development Trends of Precision Teaching

With the continuous advancement of big data technology, the future of precision teaching will rely on the deep integration of artificial intelligence (AI) and big data. AI technologies, particularly in the automation of learning behavior analysis and the automatic generation of personalized teaching strategies, show great promise. By combining big data with AI, precision teaching can more efficiently identify student learning patterns, automatically generate the most suitable learning paths and strategies, and implement fully automated teaching interventions and resource recommendations. This not only alleviates the data analysis burden on teachers but also significantly enhances the precision and personalization of teaching.

Data-driven intelligent education systems are poised to become the mainstream of future education. By integrating learning behavior data, AI algorithms, and smart teaching tools, these systems will achieve comprehensive monitoring of the learning process, real-time feedback, and dynamic intervention. This will transform the role of teachers into learning facilitators, significantly enhancing teaching efficiency, expanding the reach of personalized teaching, and promoting educational equity and improved learning outcomes.

However, as precision teaching rapidly evolves, issues of data ethics and privacy protection have become increasingly prominent. Educational institutions must ensure legal and compliant data usage during technological applications, establishing comprehensive privacy protection mechanisms to prevent data misuse. In the future, as legal regulations improve and technology advances, a better balance between data ethics and privacy protection will be achieved, providing safeguards for the healthy development of precision teaching.

Conclusion

This paper presents personalized teaching optimization strategies based on learning data through the study of big data-driven student learning behavior analysis and precision teaching. The analysis indicates that real-time monitoring, dynamic adjustments, and intelligent feedback of learning behavior data can significantly enhance teaching efficiency and learning outcomes. In the future, with the continuous development of artificial intelligence technology, precision teaching will achieve higher levels of automation and intelligence, automatically generating learning paths and teaching strategies through comprehensive learning data analysis while facilitating real-time interventions and personalized resource recommendations. However, as the application of data deepens, issues related to data privacy and ethics will emerge as urgent challenges within the education sector. Ensuring data security and privacy protection will be a key issue in the future development of precision teaching. As technology advances and legal regulations improve, the education sector will be better positioned to balance data usage with privacy protection, thereby promoting educational equity and quality improvement.

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