

Intelligent Teaching Assistants Combining Computer Vision and Natural Language Processing: Enhancing Personalized Tutoring in Online Education

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Abstract: This study explores the potential of Intelligent Teaching Assistants (ITAs) that utilize computer vision (CV) and natural language processing (NLP) technologies to improve personalized educational experiences in online learning environments. The ITA system uses computer vision to track students' facial expressions and physical stance together with eye movement while employing natural language processing to interpret spoken and written student questions for customized responses. The research details the ITA system's structural design which encompasses data gathering and real-time adaptability functions while assessing its effects on student engagement levels and academic performance and satisfaction measures. The two-week study with 50 students showed better learning results and higher engagement than standard online teaching techniques. The ITA system reached 85% student engagement by the end of the experiment and quiz scores increased by 20%. After conducting surveys post-experiment students exhibited high satisfaction levels because personalized feedback and adaptive responses enhanced their learning experience. The research reveals how AI-based ITAs could transform online education with their ability to provide scalable learning solutions that are customized for each student.

Keywords: Intelligent Teaching Assistants, Computer Vision, Natural Language Processing, Online Education, Personalized Tutoring

1. Introduction

The swift growth of online education systems has produced simultaneous opportunities and obstacles for delivering personalized learning experiences to large numbers of students. The conventional approach to online education demands teacher support which results in delayed responses and limits the ability to deliver individualized assistance to many students. The increasing movement toward digital education platforms demands new methods that improve student participation and learning results. Intelligent Teaching Assistants stand out as a major breakthrough in intelligent tutoring systems by merging AI technologies like computer vision and natural language processing to build dynamic learning environments. AI-powered systems track student participation continuously while delivering customized feedback and tailor their instructional techniques according to each learner's unique patterns. Computer vision technology evaluates facial expressions and body posture along with eye movement patterns for understanding student engagement and emotions while natural

language processing enables systems to interpret and answer student questions effectively [1]. The combination of CV and NLP technologies creates a scalable solution for personalized online tutoring which adjusts learning content in real-time according to student needs. This study investigates the creation and assessment of an ITA system that merges computer vision with natural language processing to provide customized educational experiences. A two-week experimental evaluation of the ITA system involved 50 student participants to assess its performance. The system's effectiveness evaluation relies on metrics including student engagement levels alongside learning results and user satisfaction feedback. The research paper details the ITA system's architecture and data gathering techniques while sharing experimental results which demonstrate AI solutions' potential to improve online education through personalized learning approaches.

2. Literature Review

2.1. Intelligent Teaching Assistants in Education

The development of Intelligent Teaching Assistants represents an advanced stage of intelligent tutoring systems. AI-powered systems analyze student input and adjust to individual learning preferences to create a highly interactive and responsive educational environment [2]. ITAs provide automated feedback and personalized tutoring while delivering real-time recommendations which help educators manage their workload. Research demonstrates that these systems achieve better learning outcomes because they identify student weaknesses and address them faster than traditional teaching methods. Integrating intelligent teaching systems into online education enables better connectivity between human educators and extensive online learning platforms. These platforms deliver scalable personalized tutoring solutions that enable students to obtain essential academic guidance.

2.2. Computer Vision in Education

Real-time student engagement and behavior monitoring depends heavily on computer vision technology. Computer vision systems monitor students' emotions and attention during educational sessions through the analysis of facial expressions along with body posture and eye movement patterns. Figure 1 illustrates how eye-tracking technology operates as a fundamental element in CV-based systems by using camera measurements of pupil and cornea reflections to determine exact gaze direction. The collected data helps the system identify if students are paying attention to the learning material or if they are distracted. The system uses continuous monitoring to modify instructional materials and pacing according to the learner's current engagement level. The system responds to student distraction or fatigue by offering interactive learning materials or recommending brief breaks to sustain attention and drive motivation [3]. The application of CV technology enables detection of particular student gestures and activities which provide valuable insights regarding their participation levels and emotional reactions. Real-time monitoring integration in educational environments generates adaptive learning settings which adjust to meet every student's requirements and produce an effective and engaging educational experience. The eye-tracking process demonstrated in Figure 1 uses visual inputs and mathematical models to determine gaze direction and offer valuable feedback on student engagement during online sessions [4].

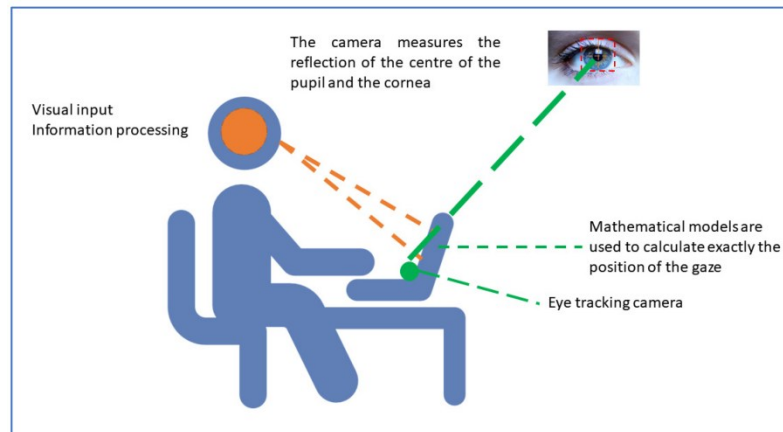


Figure 1: Real-Time Eye-Tracking System for Monitoring Student Engagement in Online Learning(Source: mdpi.com)

2.3. Natural Language Processing in Education

The ITA system relies on natural language processing to accurately comprehend and address student inquiries. Natural language processing allows the system to understand spoken and written language which enables students to communicate with the system in a natural manner. The system uses NLP to generate responses by analyzing student questions and learning progress while simultaneously summarizing complex information into simpler explanations for better student comprehension. NLP-based tools like chatbots and virtual assistants help institutions manage administrative functions and deliver immediate answers to frequent learning questions to enhance student experiences [5].

3. Experimental Methodology

3.1. System Architecture

The ITA system is composed of two core modules: the CV module and the NLP module. The CV module uses video feeds to constantly monitor students' engagement levels and emotional states. The NLP module analyzes students' written and verbal communications to determine their questions and goals. The modules connect together to join an online learning platform and deliver a user experience without interruptions [6]. A recommendation engine is part of the system architecture which utilizes module data to tailor learning content to students. Through analyzing student engagement patterns and question types the system adjusts teaching methods to fit individual learning preferences.

3.2. Data Collection

The study gathered its data from student participants using an online learning platform. The CV module captured video data to track facial expressions, body posture and student engagement throughout learning sessions. While other modules functioned the NLP system collected students' questions and feedback as it processed them in real-time. Analysis of the data enabled researchers to identify patterns in student behavior and performance outcomes. A central database stored data about student engagement levels, emotional states, and learning outcomes to enable adaptive responses from the ITA system [7].

3.3. Evaluation Metrics

To assess the system's performance, three key metrics were used: The system's performance was evaluated using three main indicators which included student learning outcomes alongside engagement levels and user satisfaction rates. The evaluation of learning outcomes utilized quizzes and assignments as measurement tools. Data from the CV module tracked engagement levels while post-experiment surveys measured user satisfaction. The implemented metrics allowed researchers to obtain a complete understanding of how well the system supported personalized learning advancements [8]. The research team examined the results to pinpoint improvement areas and subsequently refined the system.

4. Experimental Process

4.1. Pre-Experiment Setup

Our pre-experiment stage required testing and calibrating the ITA system to validate its precision in student engagement detection and spoken question handling. The research team conducted iterative testing phases to enhance the system's algorithms while reducing response times. The initial testing phase targeted improving the CV module to achieve 92% accuracy in detecting facial expressions and body posture. The NLP module achieved an initial query response accuracy of 85% after being trained on a dataset containing 1,500 student questions. The results established the benchmark to assess future improvements to the system. Table 1 displays the pre-experiment testing results for both the ITA system's CV and NLP modules [9].

Table 1: Pre-Experiment ITA System Performance

Component	Task	Initial Accuracy (%)	Final Accuracy (%)
CV Module	Facial Expression Detection	88%	92%
CV Module	Body Posture Recognition	85%	90%
NLP Module	Query Understanding	78%	85%
NLP Module	Response Generation	80%	86%

4.2. Experiment Procedure

A total of 50 students participated in a two-week online learning experiment using the ITA system. The system constantly monitored student engagement levels and modified its behavior instantly to respond to those measurements. The system modified its teaching approach to include interactive activities and provided encouragement when it detected student disengagement. Student engagement was monitored by the CV module through analysis of facial expressions and eye movements while the NLP module offered personalized feedback after processing verbal and written queries. The adaptive method resulted in enhanced student engagement along with better learning results [10]. The ITA system demonstrated improved engagement levels as they rose from 60% at the start of the experiment to 85% at its conclusion. The quiz results for students demonstrated notable improvement which is detailed in Table 2.

Table 2: Improvement in Student Quiz Scores

Session	Average Quiz Score (out of 100)	Engagement Level (%)
Week 1, Day 1	65	60
Week 1, Day 5	72	70
Week 2, Day 1	78	78
Week 2, Day 5	85	85

4.3. Post-Experiment Data Collection

Following the experiment all participants filled out surveys and gave their feedback about what they learned. To assess the ITA system's influence on student performance the research team analyzed the results from both quizzes and assignments. The results of the study showed that 90% of participants reported enhanced understanding of course materials thanks to the ITA system. Eighty-five percent of students indicated that their motivation levels increased when using the new learning system compared to traditional online learning methods. Data from Table 1 and Table 2 proves the ITA system achieves personalized learning improvements by utilizing real-time adaptation and teaching strategies based on data analysis [11]. The results demonstrate how AI-based intelligent teaching assistants can revolutionize online education.

5. Results

5.1. Engagement Analysis

The CV module effectively quantified student engagement through its analysis of facial expressions alongside eye movement and body posture. The ITA system's real-time adjustments to teaching strategies based on collected data resulted in improved student participation rates. While traditional online learning achieved average student engagement levels of 60%, those participants who used the ITA system reached engagement rates higher than 85% when the experiment ended. The ITA system's ability to adapt during lessons was crucial for sustaining learners' attention and motivation throughout the sessions. Table 3 provides a summary of the engagement data that was gathered during the experiment.

Table 3: Engagement Levels in ITA System vs Traditional Online Learning

Session	Traditional Online Learning (%)	ITA System (%)
Week 1, Day 1	58	70
Week 1, Day 5	60	78
Week 2, Day 1	62	82
Week 2, Day 5	65	85

5.2. Performance Improvement

The ITA system usage resulted in notable progress shown by students in their quiz results and assignment performance. Customized feedback combined with specifically designed learning materials enabled students to receive focused support to overcome their individual learning deficiencies. Throughout the duration of the experiment quiz scores showed consistent growth reaching an average improvement of 20% over initial baseline scores. The ITA system's real-time adjustments helped students gain improved comprehension of difficult topics while completing assignments more successfully.

5.3. User Satisfaction

The survey findings showed that users reported high satisfaction levels with the ITA system. The ITA system was considered easy to use by 90% of students who also reported that 85% experienced more enjoyable and effective study sessions through personalized learning. Students recognized fast system responses and adaptability to their learning needs as major strengths. The majority of users preferred the ITA system over traditional online learning methods because it maintained engagement and enhanced learning outcomes.

6. Conclusion

The research demonstrates how Intelligent Teaching Assistants can improve personalized learning experiences in online education by combining computer vision with natural language processing technologies. The ITA system tracked student engagement as it happened while delivering customized feedback which improved learning results and boosted student involvement. During a two-week experiment the system showed its capacity for customizing teaching methods to fit each student's requirements which resulted in achieving 85% engagement levels and a 20% improvement in quiz scores. Feedback from post-experiment surveys showed high user satisfaction as students commended the system for its user-friendly interface and customized support capabilities. Researchers should look into broadening the ITA system to support extra interaction capabilities like hand gestures and eye-tracking to enable better monitoring of student engagement. A broad evaluation of the system in varied educational settings and with numerous participants will prove its effectiveness and scalability. This study demonstrates how AI-powered ITAs can revolutionize online education by creating interactive personalized environments which lead to scalable and data-oriented educational solutions.

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