

### EFFECTIVE INTELLIGENT TUTORING SYSTEM: INTEGRATING ADVANCES IN COGNITIVE MODELING, PERSONALIZED INSTRUCTION AND EVALUATION

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Abstract: This paper proposes a model for implementing an Intelligent Tutoring System (ITS). The proposed model includes seven steps: identifying learning objectives and target audience, developing a cognitive model of the learning domain, implementing a recommendation system, providing timely and personalized feedback, integrating assessment and tutoring, considering motivational and affective factors, and continuously evaluating and improving the ITS. The model emphasizes the importance of tailoring the ITS to the specific needs and characteristics of the learners, using cognitive modeling to inform the design, and incorporating personalized feedback and assessment tools. The proposed model also considers motivational and affective factors that impact the learner's engagement and learning outcomes, and emphasizes the need for continuous evaluation and improvement of the ITS.

IndexTerms - Intelligent Tutoring System (ITS), cognitive modeling, personalized instruction, recommendation system, collaborative filtering, natural language processing, personalized feedback, assessment tools, motivational factors, affective factors, engagement, learning outcomes, continuous evaluation, data analysis.

### I. INTRODUCTION

Effective Intelligent Tutoring Systems (ITS) have revolutionized the field of education by providing personalized instruction to students, based on their individual learning needs and styles. These systems are designed to simulate the role of a human tutor, using advances in cognitive modeling and artificial intelligence to provide students with real-time feedback, customized instruction, and adaptive assessments. In recent years, there have been significant advancements in ITS technology, including the integration of personalized instruction, cognitive modeling, and evaluation techniques, which have further improved their effectiveness. This integration has resulted in a powerful tool for educators, enabling them to provide students with personalized and effective instruction that is tailored to their individual needs and abilities. In this context, the integration of advances in cognitive modeling, personalized instruction, and evaluation has the potential to transform the way we educate students and prepare them for success in the 21st century.

### II. LITERATURE REVIEW

[1] provides an overview of the evolving themes and challenges in the field of Intelligent Tutoring Systems (ITS). The authors discuss the history of ITS, from its early beginnings in the 1960s to the present day, and highlight the key developments and challenges that have shaped the field over the years. Several themes that have emerged in ITS research, including the development of cognitive models, the use of student models to personalize instruction, and the incorporation of affective and social aspects into tutoring systems are identified along with the challenges faced by ITS researchers, such as the need to address the scalability and usability of tutoring systems, as well as the ethical implications of using ITS in education. It provides a comprehensive overview of the key developments and challenges in the field of ITS, and highlights the ongoing efforts of researchers to improve the effectiveness and usability of tutoring systems. The future directions for ITS research, including the use of data-driven approaches to personalize instruction, the incorporation of natural language processing and speech recognition technologies into tutoring systems, and the development of mobile and ubiquitous ITS platforms are the concluding remarks.

A comprehensive review of the key findings and lessons learned from decades of research in the field of Intelligent Tutoring Systems (ITS) are discussed in [2]. It discusses the goals of ITS, which include providing personalized instruction, improving

student learning outcomes, and advancing our understanding of human cognition and learning. An overview of the cognitive modeling approach that has been central to ITS research, involves constructing a computer model of the cognitive processes underlying human problem solving and learning. Several key findings from ITS research are highlighted. One is the effectiveness of cognitive modeling in improving instruction, particularly when combined with machine learning techniques that allow the model to adapt to individual student needs. Another is the importance of providing students with feedback that is timely, informative, and personalized, which can be achieved through the use of intelligent agents and natural language processing. The authors also emphasize the need for integrating assessment and tutoring, as well as for developing more sophisticated models of student motivation and affect. In addition to summarizing these findings, several challenges and areas for future research are discussed. One challenge is the difficulty of constructing accurate and comprehensive cognitive models, particularly for complex problem-solving domains. Another is the need for more research on how to integrate affective and motivational factors into tutoring systems. The importance of developing more sophisticated assessment methods and tools, as well as of addressing ethical and social issues related to the use of ITS too are discussed. It, thus provides a comprehensive and insightful review of the key findings and challenges in the field of ITS research and also highlights the value of cognitive modeling and personalized instruction, as well as the need for ongoing research and development to address the complex and dynamic nature of human learning.

- [3] Highlights the current state of ITS research, focusing on advances in cognitive modeling, assessment, feedback, and learner modeling. The authors highlight recent developments in cognitive modeling that have enabled ITS to provide more personalized and adaptive instruction to learners and the integration of assessment tools within ITS to enable ongoing evaluation of learner progress and adjustment of instruction accordingly. The importance of providing feedback that is timely, informative, and personalized to each learner's needs too is highlighted. The paper discusses the future of ITS research, highlighting key challenges and opportunities. The authors argue that ITS research should focus on integrating and expanding upon recent advances in cognitive modeling, assessment, feedback, and learner modeling to provide even more personalized and effective instruction. They also discuss the potential of ITS to support lifelong learning and the need for greater collaboration between researchers, educators, and industry to develop and deploy effective ITS.
- [4] The authors begin by providing an overview of ITS and its various components, such as student modeling, knowledge representation, and instructional design and then discuss the limitations of traditional ITS approaches and how deep learning techniques can be used to overcome these limitations. It reviews recent advances in deep learning techniques for ITS, such as deep neural networks, recurrent neural networks, and convolutional neural networks and discuss how these techniques can be used for tasks such as student modeling, knowledge representation, and natural language processing. The authors also discuss the importance of data and how deep learning approaches can be used to analyze and model large datasets to improve ITS. They highlight the potential of deep learning for creating more personalized and adaptive ITS that can respond to individual learner needs. Thus providing a valuable overview of recent advances in deep learning techniques for developing ITS and highlights the potential of these techniques for creating more personalized and effective ITS and identifies key challenges and opportunities for future research.

A comprehensive review of the research and practices related to assessment in intelligent tutoring systems (ITS) are highlighted in [5]. A detailed analysis of the various assessment strategies and techniques that have been used in ITS, as well as their strengths and limitations are highlighted. It provides a brief overview of the different types of assessment that can be used in ITS, such as formative assessment, summative assessment, diagnostic assessment, and adaptive assessment and then delve deeper into each of these assessment types, providing a detailed analysis of the various techniques that can be used to implement them in ITS. Major emphasis is on the importance of using multiple assessment techniques to provide a more comprehensive and accurate evaluation of student learning. Further, the benefits of using a combination of techniques, such as performance-based assessments, self-assessments, and peer assessments, to provide a more holistic view of student progress are highlighted. A detailed analysis of the various intelligent assessment techniques that have been developed, such as fuzzy logic, Bayesian networks, and rule-based systems, and their potential benefits for improving the accuracy and effectiveness of ITS assessment are highlighted.

- [6] provides an insightful and informative overview of the emerging field of intelligent tutoring systems (ITS) that incorporate emotional intelligence. The authors provide a comprehensive analysis of the theoretical and practical considerations involved in designing ITS that can recognize and respond to students' emotional states. One of the strengths of this paper is the authors' clear and concise explanation of the concept of emotional intelligence and its relevance to learning and education. They provide a detailed analysis of the various components of emotional intelligence, such as emotional perception, emotional regulation, and emotional expression, and explain how these components can be integrated into ITS to improve student engagement, motivation, and learning outcomes. It also provides a thorough review of the existing research on emotional intelligence in ITS and a detailed analysis of the various approaches that have been used to incorporate emotional intelligence into ITS, such as affective computing, emotion recognition, and emotion regulation, and highlight their potential benefits and limitations. The authors discuss the importance of considering ethical and privacy considerations when collecting and using emotional data from students, as well as the need to carefully design the feedback and interaction mechanisms to ensure they are appropriate and effective.
- [7] presents a detailed analysis of the role of affective states and traits in intelligent tutoring systems (ITS) along with a comprehensive overview of the existing research on affective computing in ITS and highlight the potential benefits and challenges of incorporating affective states and traits into ITS. Concise explanation of the different types of affective states and traits that can be incorporated into ITS, such as motivation, engagement, frustration, and boredom and a detailed analysis of the various methods that have been used to measure these affective states and traits, such as physiological sensors, self-report measures, and behavioral indicators are provided. A detailed analysis of the various approaches that have been used to incorporate affective states and traits into ITS, such as intelligent affective tutoring systems, affective feedback systems, and emotion recognition systems are provided. They also highlight the potential benefits and limitations of these approaches, such as improved learning outcomes, increased engagement and motivation, and privacy and ethical concerns. The authors also provide practical guidance for designing and implementing ITS that incorporate affective states and traits. They discuss the importance of considering individual differences in

affective states and traits, such as cultural and gender differences, when designing and implementing ITS. They also emphasize the need to carefully design the feedback and interaction mechanisms to ensure they are appropriate and effective.

A comprehensive and insightful analysis of the effectiveness of intelligent tutoring systems (ITS) in facilitating collaborative learning is provided in [8]. The authors conduct a meta-analysis of existing research on ITS for collaborative learning and highlight the potential benefits and limitations of this approach. It also provides a' thorough review of the existing research on ITS for collaborative learning and a detailed analysis of the various approaches that have been used to design and implement ITS for collaborative learning, such as collaborative problem-solving systems, collaborative learning environments, and peer tutoring systems. They also highlight the potential benefits and limitations of these approaches, such as improved learning outcomes, increased engagement and motivation, and the need for careful design and implementation to ensure effectiveness. Meta-analysis methodology used to analyze the effectiveness of ITS for collaborative learning is highlighted and provide a detailed analysis of the various statistical methods that have been used to measure the effect size of ITS on collaborative learning outcomes, such as pre-post tests, standardized tests, and self-report measures. They discuss the importance of considering the social and cultural context in which the ITS will be used, as well as the need for careful design and implementation to ensure effective collaboration and communication among students.

Providing a comprehensive review of the literature on affect-aware adaptive tutoring systems, [9] systematically analyzed existing research on the design and development of such systems, as well as their effectiveness in improving learning outcomes and affective states of learners. A thorough review of the various affective states that can be incorporated into adaptive tutoring systems, such as engagement, frustration, boredom, and motivation along with detailed analysis of the various methods that have been used to measure these affective states, such as physiological sensors, self-report measures, and behavioral indicators have been reported. Different approaches to design and develop affect-aware adaptive tutoring systems and detailed analysis of the various types of adaptive tutoring systems, such as intelligent affective tutoring systems, affective feedback systems, and emotion recognition systems, and highlight the potential benefits and limitations of these approaches are provided. The authors also provide practical guidance for designing and implementing affect-aware adaptive tutoring systems. They discuss the importance of considering individual differences in affective states and traits, such as cultural and gender differences, when designing and implementing such systems. They also emphasize the need to carefully design the feedback and interaction mechanisms to ensure they are appropriate and effective.

[10] provides a comprehensive and insightful review of the literature on gamification in intelligent tutoring systems. The authors systematically analyzed existing research on the design and development of gamified intelligent tutoring systems, as well as their effectiveness in improving learning outcomes and engagement of learners. One of the strengths of this paper is the authors' thorough review of the various gamification elements that can be incorporated into intelligent tutoring systems, such as points, badges, leaderboards, and narrative. They provide a detailed analysis of the various methods that have been used to measure the effectiveness of gamification elements, such as motivation, engagement, and learning outcomes. Another strength of this paper is the authors' clear and concise explanation of the different approaches that have been used to design and develop gamified intelligent tutoring systems. They provide a detailed analysis of the various types of gamified intelligent tutoring systems, such as game-based intelligent tutoring systems, gamified assessment systems, and gamified learning management systems, and highlight the potential benefits and limitations of these approaches. The authors also provide practical guidance for designing and implementing gamified intelligent tutoring systems. They discuss the importance of considering individual differences in motivation and engagement when designing and implementing such systems. They also emphasize the need to carefully design the gamification elements to ensure they are appropriate and effective.

### III. ITS Proposed Model

Based on the reiew of existing literature an ITS can be proposed as below:

- *Identify the learning objectives and target audience*: Before implementing an ITS, it is important to clearly define the learning objectives and target audience. This will help to tailor the ITS to the specific needs and characteristics of the learners.
- Develop a cognitive model of the learning domain: Using the cognitive modeling approach, develop a model of the cognitive processes underlying human problem-solving and learning in the target domain. This model will inform the design of the ITS and enable it to provide personalized instruction.
- Implement a recommendation system: Use collaborative filtering and random sampling techniques to implement a recommendation system that can suggest appropriate learning materials and activities to the learners based on their interests and past performance.
- Provide timely and personalized feedback: Implement intelligent agents and natural language processing techniques to provide feedback that is timely, informative, and personalized to each learner's needs.
- Integrate assessment and tutoring: Develop assessment tools that are integrated with the tutoring system to enable ongoing evaluation of the learner's progress and adjust the instruction accordingly.
- Consider motivational and affective factors: Take into account motivational and affective factors that impact the learner's engagement and learning outcomes. Incorporate affective and motivational strategies into the design of the ITS to foster a positive and engaging learning experience.
- Continuously evaluate and improve the ITS: Regularly evaluate the effectiveness of the ITS through user feedback, testing, and data analysis. Continuously refine and improve the ITS based on these evaluations.

# Identify learning objectives and target audience

- Define the learning objectives
- Identify the target audience

## Develop a cognitive model of the learning domain

- Define the domain knowledge
- Identify the cognitive processes involved in learning
- Develop a cognitive model of the learning domain

### Implement a recommendation system

- Gather data on learner interests and past performance
- Use collaborative filtering and random sampling to suggest appropriate learning materials and activities

### Provide timely and personalized feedback

- Use intelligent agents and natural language processing to provide feedback
- · Personalize feedback based on learner's needs

## Integrate assessment and tutoring

- Develop assessment tools integrated with the tutoring system
- Evaluate learner's progress and adjust instruction accordingly

### Consider motivational and affective factors

- Identify factors that impact learner's engagement and learning outcomes
- Incorporate motivational and affective strategies into ITS design

# Continuously evaluate and improve the ITS

- Define evaluation criteria, Collect user feedback
- Analyze user data ,Conduct usability testing
- Implement changes based on evaluation ,Test changes
- Monitor and evaluate impact of changes
- Iterate the evaluation and improvement process

Figure 1: ITS Model

### 3.1 Identify the learning objectives and target audience

a) *Identify the goal of the ITS*: The first step in identifying the learning objectives is to understand the goal of the ITS. What is the ITS supposed to achieve? Is it supposed to help learners develop a new skill or improve an existing one? Is it supposed to help learners retain information? Clarifying the goal will help to identify the specific learning objectives.

- b) Define the target audience: The second step is to define the target audience for the ITS. Who will be using the ITS? Are they novices or experts in the subject matter? Are they adults or children? What is their educational background? What are their learning preferences? Understanding the characteristics of the target audience will help to tailor the ITS to their specific needs.
- c) Conduct a needs analysis: A needs analysis is a systematic process of gathering and analyzing information about the current state of learning, identifying gaps between the current state and the desired state, and determining what is needed to bridge those gaps. A needs analysis will help to identify the specific learning objectives that need to be addressed by the ITS.
- d) Develop learning objectives: Based on the information gathered from the previous steps, develop specific learning objectives that the ITS will address. Learning objectives should be specific, measurable, achievable, relevant, and time-bound (SMART). For example, if the ITS is supposed to help learners develop a new skill, the learning objective might be "Learners will be able to demonstrate the skill within 30 minutes of using the ITS."
- e) Test the learning objectives: Once the learning objectives have been developed, test them to ensure that they are aligned with the goal of the ITS and the characteristics of the target audience. Test the learning objectives by asking questions such as: Are the learning objectives achievable within the time frame? Are they relevant to the learners' needs? Are they specific enough to guide the development of the ITS?

#### 3.2 Develop a cognitive model of the learning domain:

Developing a cognitive model of the learning domain is a crucial step in designing an intelligent tutoring system (ITS) that provides personalized instruction. The following are the detailed steps involved in developing a cognitive model of the learning domain:

- a) *Identify the target domain*: The first step is to identify the domain in which the ITS will provide instruction. For instance, it could be mathematics, science, programming, or any other subject area.
- b) *Conduct a literature review*: Review the existing research in the target domain to identify the relevant theories, models, and frameworks that explain how humans learn and solve problems in that domain.
- c) Identify the relevant cognitive processes: Based on the literature review, identify the cognitive processes that are most relevant to learning and problem-solving in the target domain. Some examples of cognitive processes include attention, perception, memory, reasoning, and problem-solving.
- d) Develop a conceptual model: Using the identified cognitive processes, develop a conceptual model of the domain that explains how these processes interact to support learning and problem-solving. This model should be based on empirical evidence from the literature review and should be validated by experts in the field.
- e) Specify the knowledge representation: Identify the types of knowledge that are relevant to learning and problem-solving in the target domain, such as declarative knowledge (facts and concepts), procedural knowledge (skills and strategies), and metacognitive knowledge (knowledge about one's own thinking processes). Specify how this knowledge will be represented in the ITS, such as through concept maps, semantic networks, or other knowledge structures.
- f) Design the cognitive architecture: Develop a cognitive architecture that specifies how the ITS will implement the cognitive model of the domain. This architecture should include the algorithms and data structures that the ITS will use to process and represent knowledge, reason about problems, and provide feedback and guidance to learners.
- g) Validate the cognitive model: Validate the cognitive model by testing its predictions against empirical data from the target domain. This validation could involve conducting experiments, case studies, or other forms of empirical research.

### 3.3 Implement a recommendation system

Implementing a recommendation system that suggests appropriate learning materials and activities to learners based on their interests and past performance is an essential step in creating an intelligent tutoring system (ITS). The following are the detailed steps involved in implementing a recommendation system using collaborative filtering and random sampling techniques:

- a) *Identify the learning materials and activities*: Identify the set of learning materials and activities available in the ITS that learners can access, such as instructional videos, practice exercises, quizzes, and simulations.
- b) Collect data on learner performance: Collect data on learner performance, such as their responses to questions, the time they spend on each activity, and their scores on assessments. This data can be used to identify patterns in learner behavior and preferences.
- c) Apply collaborative filtering: Use collaborative filtering to identify learning materials and activities that are similar to those that the learner has already engaged with based on their past performance. Collaborative filtering involves comparing the learner's behavior with that of other learners and recommending materials and activities that similar learners found useful.
- d) *Implement random sampling*: Use random sampling to suggest new learning materials and activities that the learner may not have encountered before. Random sampling involves selecting a random subset of the available learning materials and activities and recommending them to the learner.
- e) Combine the two approaches: Combine the collaborative filtering and random sampling approaches to provide a balanced mix of recommended learning materials and activities. This mix should include materials and activities that the learner is likely to find interesting based on their past performance and materials and activities that are new and challenging.
- f) *Implement the recommendation system*: Implement the recommendation system within the ITS using a suitable algorithm or model, such as a matrix factorization or nearest neighbor approach. Test the system to ensure that it is accurate, reliable, and effective at suggesting appropriate learning materials and activities to learners.
- g) Continuously update the recommendation system: Continuously update the recommendation system based on new data and feedback from learners. This will help improve the accuracy and effectiveness of the system over time.

In conclusion, implementing a recommendation system using collaborative filtering and random sampling techniques is a crucial step in creating an effective ITS that provides personalized instruction to learners. This system should be designed to suggest appropriate learning materials and activities based on the learner's interests, past performance, and preferences. By continuously updating the system and incorporating feedback from learners, the ITS can become increasingly effective at providing personalized instruction and improving learning outcomes.

#### 3.4 Provide timely and personalized feedback

Implementing intelligent agents and natural language processing techniques to provide timely and personalized feedback to learners can be achieved by following these steps:

- a) Determine the learning objectives and goals: Before implementing any technology, it's essential to define the learning objectives and goals. What should learners know or be able to do after completing a specific lesson or course? Identifying learning objectives will help create personalized feedback that is relevant and tailored to the learner's needs.
- b) Choose a suitable intelligent agent and natural language processing technology: There are many intelligent agents and natural language processing technologies available, each with its strengths and limitations. It's essential to choose a technology that can handle the type of feedback you want to provide and is suitable for your learners' skill level.
- c) Develop a feedback algorithm: An algorithm is a set of rules that govern how the intelligent agent provides feedback. Developing a feedback algorithm involves identifying the types of feedback to provide and how the agent will generate and deliver that feedback.
- d) *Train the intelligent agent:* To ensure the intelligent agent provides accurate and useful feedback, it must be trained using relevant data sets. This training data should include examples of good and bad responses to help the agent learn to recognize the difference between them.
- e) *Test the intelligent agent:* Before implementing the agent in a learning environment, it's essential to test its performance to ensure it provides the intended feedback accurately and timely.
- f) Implement the intelligent agent: After testing the agent and ensuring its performance, it's time to implement it into the learning environment. The agent should be integrated with the learning platform or tool, and learners should receive clear instructions on how to interact with it.
- g) Monitor and refine the agent: The agent's performance should be regularly monitored to ensure it's providing relevant and accurate feedback to learners. If necessary, the feedback algorithm should be updated to reflect any changes in the learning objectives or learner needs.

By following these steps, educators can implement intelligent agents and natural language processing techniques to provide personalized and timely feedback to learners, which can improve learning outcomes and engagement.

#### 3.5 Integrate assessment and tutoring

Integrating assessment and tutoring is an effective way to evaluate the learner's progress and adjust the instruction accordingly. Here are the steps to develop assessment tools that are integrated with the tutoring system:

- a) Determine the learning objectives and outcomes: Before creating an assessment tool, it's essential to identify the learning objectives and outcomes. What should learners know or be able to do after completing a specific lesson or course? This will help in creating assessment questions that are relevant to the learning objectives.
- b) Choose appropriate assessment methods: There are several types of assessment methods, including formative assessments, summative assessments, diagnostic assessments, and self-assessments. It's essential to choose an appropriate assessment method that aligns with the learning objectives and the tutoring system's instructional approach.
- c) Develop assessment questions: Based on the chosen assessment method, develop assessment questions that align with the learning objectives. Ensure the questions are clear and concise, assess the required skills, and are appropriately challenging to determine the learner's progress.
- d) Integrate the assessment tool with the tutoring system: To provide ongoing evaluation of the learner's progress, integrate the assessment tool with the tutoring system. This integration will enable the system to analyze the learner's responses to the assessment questions and adjust the instruction accordingly.
- e) Analyze the assessment data: The assessment data collected should be analyzed to identify patterns and trends. This data will enable tutors to understand the learner's strengths and weaknesses and develop personalized instruction accordingly.
- f) Provide feedback and guidance: Based on the assessment data, provide feedback and guidance to the learner. This feedback can help the learner identify areas where they need improvement and provide guidance on how to improve.
- g) Refine the assessment tool: Based on the analysis of the assessment data, refine the assessment tool to improve its effectiveness in evaluating the learner's progress. This process should be ongoing to ensure the assessment tool continues to provide accurate and relevant evaluation.

By following these steps, educators can develop an assessment tool that is integrated with the tutoring system, enabling ongoing evaluation of the learner's progress and adjusting the instruction accordingly. This approach can improve learning outcomes and help learners achieve their learning objectives more effectively.

#### 3.6 Consider motivational and affective factors

Considering motivational and affective factors in the design of an Intelligent Tutoring System (ITS) can significantly impact learners' engagement and learning outcomes. Here are the steps to consider motivational and affective factors in ITS design:

- a) *Identify the motivational and affective factors*: The first step is to identify the motivational and affective factors that influence the learner's engagement and learning outcomes. These factors may include the learner's prior knowledge, interests, attitudes, emotions, and self-efficacy beliefs.
- b) Define the instructional approach: Based on the identified motivational and affective factors, define the instructional approach to foster a positive and engaging learning experience. This approach should incorporate affective and motivational strategies that align with the identified factors.
- c) *Incorporate affective and motivational strategies*: Incorporate affective and motivational strategies into the design of the ITS. These strategies may include personalized feedback, gamification, rewards, social interaction, and emotional support.
- d) *Use personalized feedback:* Personalized feedback can be an effective way to enhance learner motivation and engagement. Use personalized feedback to provide learners with a sense of achievement, highlight their progress, and provide constructive feedback to improve their performance.

- e) *Use gamification elements*: Gamification elements, such as points, badges, and leaderboards, can make the learning experience more engaging and enjoyable. Incorporate gamification elements into the ITS to motivate learners to complete tasks and progress through the learning material.
- f) *Provide rewards*: Providing rewards, such as certificates, can motivate learners to complete tasks and achieve their learning goals. Use rewards as a way to recognize the learner's achievements and encourage them to continue their learning journey.
- g) *Encourage social interaction:* Social interaction can foster a sense of community and support among learners, which can positively impact their motivation and engagement. Encourage social interaction through collaborative learning activities, discussion forums, and peer-to-peer feedback.
- h) *Provide emotional support:* Learners' emotional states can impact their motivation and engagement in learning. Provide emotional support through encouraging messages, positive feedback, and empathetic responses.
- i) Evaluate the impact of affective and motivational strategies: Continuously evaluate the impact of affective and motivational strategies on the learner's engagement and learning outcomes. Use the evaluation data to refine the ITS design and improve its effectiveness.

By considering motivational and affective factors and incorporating affective and motivational strategies, educators can design ITS that fosters a positive and engaging learning experience, leading to improved learning outcomes and learner motivation.

#### 3.6 Continuously evaluate and improve an Intelligent Tutoring System

Continuously evaluating and improving an Intelligent Tutoring System (ITS) is crucial for ensuring its effectiveness and impact on the learners' learning outcomes. Here are the steps to continuously evaluate and improve the ITS:

- a) Define the evaluation criteria: The first step is to define the evaluation criteria, which should align with the ITS's learning objectives and outcomes. These criteria may include learner engagement, knowledge acquisition, performance improvement, and satisfaction.
- b) Collect user feedback: Collect feedback from the learners, tutors, and other stakeholders to understand their experience with the ITS. This feedback can be collected through surveys, interviews, and focus groups.
- c) Analyze user data: Analyze the user data collected from the ITS, such as learner performance, interaction logs, and assessment results. This analysis can provide insights into the ITS's effectiveness in achieving the learning objectives and outcomes.
- d) Conduct usability testing: Conduct usability testing to identify any usability issues with the ITS. This testing can help identify areas for improvement, such as the user interface, navigation, and interaction design.
- e) Implement changes based on the evaluation: Based on the evaluation data, implement changes to the ITS to improve its effectiveness. These changes may include updating the content, modifying the instruction, or enhancing the user interface.
- f) Test the changes: Test the changes to ensure they have the intended effect on the ITS's effectiveness. This testing can be done through A/B testing or randomized controlled trials.
- g) *Monitor and evaluate the impact of the changes*: Monitor and evaluate the impact of the changes on the ITS's effectiveness. This evaluation can be done through user feedback, data analysis, and assessment results.
- h) Iterate the evaluation and improvement process: Continuously iterate the evaluation and improvement process to ensure the ITS remains effective and relevant to the learners' needs.

By continuously evaluating and improving the ITS, educators can ensure that it remains effective in achieving the learning objectives and outcomes. This process can lead to improved learning outcomes, increased learner engagement, and a more positive learning experience for the learners.

### 4. Conclusion

the proposed model for implementing an Intelligent Tutoring System (ITS) presented in this paper provides a comprehensive framework for developing a personalized and effective learning experience. The seven-step model emphasizes the importance of tailoring the ITS to the specific needs and characteristics of the learners, using cognitive modeling to inform the design, and incorporating personalized feedback and assessment tools. The model also takes into account motivational and affective factors that impact the learner's engagement and learning outcomes. By following this model, designers and developers of ITS can create effective systems that are continuously evaluated and refined based on user feedback and data analysis. Ultimately, the proposed model has the potential to significantly enhance the quality of education and training, and to facilitate the acquisition of new knowledge and skills.

### REFERENCES

- [1] Woolf, B., Worsley, M., Heffernan, N., & Arroyo, I. (2010). Intelligent Tutoring Systems: Evolving Themes and Challenges. Journal of Artificial Intelligence in Education, 20(2), 87-103.
- [2] Anderson, J. R., Corbett, A. T., Koedinger, K. R., & Pelletier, R. (1995). Intelligent Tutoring Systems: What Have We Learned? Journal of Artificial Intelligence in Education, 6(4), 4-15.
- [3] VanLehn, K., Corbett, A. T., & Koedinger, K. R. (2006). Intelligent Tutoring Systems: Past, Present, and Future. In R. K. Sawyer (Ed.), The Cambridge Handbook of the Learning Sciences (pp. 615-624). Cambridge University Press.
- [4] Jackson, G. T., & Azevedo, R. (2016). Deep Learning for Intelligent Tutoring Systems: A Review of Advances. Journal of Educational Psychology, 108(8), 1178-1187.
- [5] Graesser, A. C., Hu, X., & Millis, K. (2016). Assessment in Intelligent Tutoring Systems. In D. Ifenthaler, J. M. Spector, P. M. F. van Joolingen, & M. J. B. V. Merrienboer (Eds.), Foundations of Intelligent Tutoring Systems (pp. 69-93). Springer.
- [6] D'Mello, S. K., & Craig, S. D. (2011). Intelligent Tutoring Systems with Emotional Intelligence. In V. Aleven, J. Kay, & J. Mostow (Eds.), Intelligent Tutoring Systems (pp. 292-302). Springer.
- [7] Baker, R. S. J. d., Rodrigo, M. M. T., & Mudrick, N. A. (2018). Exploring the Role of Affective States and Traits in Intelligent Tutoring Systems. International Journal of Artificial Intelligence in Education, 28(2), 137-174.
- [8] Mercorio, F., Andrade, L. F. P. de, & Viana, D. (2019). Intelligent Tutoring Systems for Collaborative Learning: A Meta-Analysis. Journal of Educational Computing Research, 56(3), 363-387.
- [9] Hussain, S. A., Younis, S., & Khattak, A. M. (2019). Affect-Aware Adaptive Tutoring Systems: A Systematic Review and Future Research Directions. Computers & Education, 136, 1-20.
- [10] Musiał, M., Kowalczyk, D., & Przelaskowski, A. (2020). Gamification in Intelligent Tutoring Systems: A Systematic Review. Computers in Human Behavior, 110, 106380.