

Understanding Student Engagement in AI–Powered Online Learning Platforms: A Narrative Review of Key Theories and Models

Manuel B. Garcia

✉ <https://orcid.org/0000-0003-2615-422X>

*FEU Institute of Technology,
Philippines*

Joanna Rosak-Szyrocka

✉ <https://orcid.org/0000-0002-5548-6787>

*Czestochowa University of Technology,
Poland*

Chai Lee Goi

✉ <https://orcid.org/0000-0003-0131-2818>

Curtin University, Sarawak, Malaysia

Ari Happonen

✉ <https://orcid.org/0000-0003-0744-1776>

LUT University, Finland

Kate Shively

✉ <https://orcid.org/0000-0002-9916-0535>

Ball State University, USA

Aras Bozkurt

✉ <https://orcid.org/0000-0002-4520-642X>

Anadolu University, Turkey

Damian Maher

✉ <https://orcid.org/0000-0002-3566-0805>

*University of Technology Sydney,
Australia*

Robertas Damaševičius

✉ <https://orcid.org/0000-0001-9990-1084>

Vytautas Magnus University, Lithuania

EXECUTIVE SUMMARY

Online learning has become fundamental to modern academic and professional development. Amidst its widespread adoption, there is increasing integration of artificial intelligence (AI) to enhance the learning experience. Understanding student engagement within these AI-powered digital platforms is crucial, as it directly

influences learning outcomes and satisfaction. This chapter provides a narrative review of key theories and models essential for analyzing engagement in virtual learning contexts. Particularly, it focuses on constructivist learning theory, social learning theory, cognitive load theory, flow theory, technology acceptance model, self-determination theory, cognitive theory of multimedia learning, and feedback intervention theory. By examining these frameworks through an epistemological lens, the chapter explores how knowledge acquisition, cognitive processing, and social learning principles interact within AI-enhanced educational contexts. The insights reported here can serve as a guide for optimizing AI to maximize student involvement and educational efficacy.

INTRODUCTION

The educational landscape has been undergoing extensive transformations. Due to recent and constant disruptions to physical classes, online learning is emerging as a dominant mode of instructional delivery. The COVID-19 pandemic, in particular, accelerated the shift toward digital education (Bozkurt et al., 2022; Ofosu-Ampong et al., 2024). This change compelled institutions worldwide to adapt quickly to emergency remote teaching and learning methods. This transition highlighted the flexibility and accessibility of online learning, making it an attractive option for students and educators alike (Bethhäuser et al., 2023). With technological advancements and increasing internet accessibility, online learning has evolved from a supplementary tool to a primary mode of education delivery. It has democratized education by providing opportunities for learners from diverse geographical locations and backgrounds, allowing them to access high-quality educational resources and instruction without the constraints of time and place. Consequently, online learning has not only become more prevalent but also more sophisticated, incorporating a range of interactive and personalized learning experiences (Iyer et al., 2022).

In online learning, student engagement is one of the critical factors for determining the effectiveness of the educational experience (Aliyu et al., 2022). Student engagement refers to the level of interest, curiosity, and participation that students exhibit in the learning process. It encompasses emotional, behavioral, and cognitive dimensions, influencing how students interact with the course content, their peers, and instructors (Reschly & Christenson, 2012). Engaged students are more likely to retain information, perform better academically, and develop essential skills such as critical thinking and problem-solving (Bond et al., 2020; Li & Xue, 2023). Fostering engagement in online learning contexts is particularly important due to the lack of physical presence, face-to-face interaction, and the natural social cues inherent in traditional classroom settings. Without these elements, students

28 more pages are available in the full version of this document, which may be purchased using the "Add to Cart" button on the publisher's webpage: www.igi-global.com/chapter/understanding-student-engagement-in-ai-powered-online-learning-platforms/367141

Related Content

Clustering Categorical Data with k-Modes

Joshua Zhexue Huang (2009). *Encyclopedia of Data Warehousing and Mining, Second Edition* (pp. 246-250).

www.irma-international.org/chapter/clustering-categorical-data-modes/10828

Biological Image Analysis via Matrix Approximation

Jieping Ye, Ravi Janardanand Sudhir Kumar (2009). *Encyclopedia of Data Warehousing and Mining, Second Edition* (pp. 166-170).

www.irma-international.org/chapter/biological-image-analysis-via-matrix/10815

Reasoning about Frequent Patterns with Negation

Marzena Kryszkiewicz (2009). *Encyclopedia of Data Warehousing and Mining, Second Edition* (pp. 1667-1674).

www.irma-international.org/chapter/reasoning-frequent-patterns-negation/11042

Facial Recognition

Rory A. Lewis and Zbigniew W. Ras (2009). *Encyclopedia of Data Warehousing and Mining, Second Edition* (pp. 857-862).

www.irma-international.org/chapter/facial-recognition/10920

Stages of Knowledge Discovery in E-Commerce Sites

Christophe Giraud-Carrier (2009). *Encyclopedia of Data Warehousing and Mining, Second Edition* (pp. 1830-1834).

www.irma-international.org/chapter/stages-knowledge-discovery-commerce-sites/11067