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Digital Learning Readiness and Student Engagement in Higher Education: A Model for Effective Educational Technology Adoption

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Abstract

Digital learning has moved from a supplementary mode of delivery to a central component of higher education, yet access to educational technology does not automatically translate into meaningful learning or sustained student engagement. This paper develops an integrated model of digital learning readiness and student engagement by synthesizing evidence across educational-technology adoption, online and blended learning, digital literacy, self-efficacy, learning analytics, learning management systems, and engagement research. The synthesis examines four key readiness dimensions—digital competence, technology self-efficacy, technology attitude and perceived usefulness, and facilitating conditions—and four major engagement dimensions—behavioral, cognitive, emotional, and social or collaborative engagement. Across the evidence, technology adoption is consistently associated with perceived usefulness, usability, quality, support, and students' confidence in using digital tools; however, engagement outcomes depend strongly on pedagogical design, interaction, feedback, and learners' ability to use technology strategically. The findings therefore support a sequential model in which digital readiness enables effective educational-technology use, while effective use transforms technological capability into meaningful engagement. The model also positions institutional support, accessibility, and equitable access as important boundary conditions shaping outcomes. The paper contributes a practical framework for institutions seeking to move beyond technology provision toward student-centred digital learning environments and identifies priorities for future research, particularly empirical validation across diverse higher education contexts.

1. Introduction

This distinction matters because technology access and technology effectiveness are not equivalent. Research on e-learning success has shown that system, information, service, educational, learner, instructor, and support quality all contribute to the success of an e-learning system, with perceived usefulness, satisfaction, and actual use forming important benefit pathways (Al-Fraihat et al., 2020). More recent evidence likewise indicates that educational-technology adoption is shaped by performance

expectancy, effort expectancy, social influence, and facilitating conditions (Feng et al., 2025). These findings suggest that adoption is not simply a function of installing platforms; it is an interaction between technology characteristics, institutional conditions, and learner capabilities.

Student engagement adds a further layer of complexity. Engagement is multidimensional and includes observable participation as well as cognitive investment, emotional involvement, and social or collaborative participation. The systematic evidence

map by Bond et al. (2020) found that digital technologies were frequently associated with behavioral, affective, and cognitive engagement, but also reported limited theoretical framing and geographic concentration in the United States and United Kingdom. Nkomo et al. (2021) similarly showed that the digital engagement literature spans different technologies and engagement constructs, making it difficult to reduce engagement to platform activity alone. More recent learning-analytics research reinforces this point: Bergdahl et al. (2024) found that higher-education learning-analytics studies overwhelmingly operationalize engagement through observable behavioral indicators such as clicks and task duration, while comparatively few studies capture multiple dimensions of engagement.

Digital readiness provides a useful bridge between adoption and engagement. In the present paper, digital learning readiness is treated as a learner-and-context construct rather than a single technical skill. It includes digital competence and literacy, confidence in using digital technologies, positive or utilitarian attitudes toward educational technologies, and access to facilitating conditions such as infrastructure and support. This broader interpretation is consistent with evidence showing that digital literacy and technology attitudes can shape digital self-efficacy, which in turn predicts multiple forms of online engagement (Getenet et al., 2024). It also aligns with research showing that institutional support, course design, usability, and technology quality affect the conditions under which digital learning can succeed (Al-Fraihat et al., 2020; Alqahtani & Rajkhan, 2020).

This study addresses that fragmentation by integrating these strands into one model. The paper asks four questions: (1) Which dimensions most consistently characterize digital learning readiness in higher education technology research? (2) How is student engagement operationalized across educational-technology studies? (3) Which technology, learner, and institutional factors are repeatedly linked to effective use and engagement? and (4) How can these findings be integrated into a practical model of effective educational-technology adoption?

The contribution is intentionally conceptual and evidence-synthesizing. Rather than reporting fabricated primary survey data, the paper maps the evidence contained in a focused review corpus and uses that synthesis to derive a model that can later be tested empirically. The proposed model treats readiness as the foundation, effective use as the proximal mechanism, and engagement as the principal learner outcome. It also positions institutional support and equitable access as boundary conditions that can strengthen or weaken the pathway.

2. Theoretical Background

2.1 Educational Technology as a Learning Ecosystem

Educational technology is best understood as a learning ecosystem rather than a collection of isolated tools. Online and blended learning research shows that the educational value of technology depends on interactions among learners, instructors, course structures, organizational support, and the technologies themselves (Martin et al., 2020; Garrison & Vaughan, 2013). This ecosystem perspective also explains why the same platform can produce different learning experiences in different institutions. A technically capable learning-management system may support meaningful collaboration in one course while functioning mainly as a document repository in another. Reich (2020) and Selwyn (2016) caution against technological determinism: technology does not transform education by its presence alone, and educational effects depend on institutional practice, pedagogy, and social context.

The ecosystem view is particularly relevant after the rapid expansion of digital learning during the pandemic. Bond et al. (2021) mapped 282 empirical studies of the first global online semester and found a strong emphasis on undergraduate perceptions and descriptive, cross-sectional designs. Haider (2021) similarly documented the rapid expansion of online platforms and the persistence of digital-equity challenges. These findings support a shift from an adoption-centered question - whether technology was used - toward a learning-centered question - whether learners were able and willing to use technology effectively under the conditions of a specific course and institution.

2.2 Digital Learning Readiness

Readiness is commonly associated with the ability and willingness to participate in technology-mediated learning, but the literature does not support a single universal readiness measure. The evidence instead points toward a bundle of learner capabilities and contextual conditions. Digital competence concerns practical skills and critical use; self-efficacy concerns confidence; attitude and perceived usefulness concern value and acceptance; and facilitating conditions concern the external resources that make use possible. Feng et al. (2025) identify performance expectancy, effort expectancy, social influence, and facilitating conditions as recurring adoption dimensions, while Getenet et al. (2024) provide direct evidence for links among digital attitudes, digital literacy, self-efficacy, and engagement.

This paper therefore defines digital learning readiness as the degree to which a learner possesses the competence, confidence, motivation, and contextual

support needed to participate effectively in technology-enhanced learning. The definition deliberately avoids equating readiness with technical skill. A learner may have high technical competence but low confidence, negative attitudes toward the course technology, or insufficient access to stable connectivity. Conversely, a learner with moderate technical competence may succeed when the platform is intuitive, the instructor provides clear guidance, and peers provide collaborative support. Readiness is consequently dynamic and relational.

2.3 Student Engagement in Digital Environments

Student engagement is treated in this study as a multidimensional learning construct. Behavioral engagement refers to participation, attendance, task completion, and observable activity. Cognitive engagement concerns effort, strategy, reflection, and deep processing. Emotional engagement includes interest, belonging, enjoyment, and perceived value. Social or collaborative engagement captures interaction with peers and instructors. Bond et al. (2020) and Nkomo et al. (2021) demonstrate that educational-technology research uses multiple engagement concepts, while Bergdahl et al. (2024) show that learning-analytics studies disproportionately capture behavioral indicators.

The multidimensional view also changes how technology success should be evaluated. A learning platform that increases login frequency may improve behavioral activity without improving cognitive engagement. A collaborative system may support social engagement while imposing cognitive overload. A gamified system may raise motivation but produce only limited gains in competence (Li et al., 2024). Therefore, the proposed model treats engagement as an outcome with multiple dimensions rather than as a single score.

2.4 Adoption, Effective Use, and Learning Value

Technology adoption theories commonly focus on acceptance and intention to use, while learning-technology research increasingly asks what happens after adoption. Al-Fraihat et al. (2020) show that e-learning success involves multiple quality levels, including system, information, service, educational, support, learner, and instructor factors. This is important because adoption can be high even when educational value is low. A platform may be widely used because it is institutionally required, not because students perceive it as useful or because the course design creates meaningful interaction.

The present model therefore separates adoption from effective use. Adoption means that a student or institution uses a technology. Effective use means that the technology is usable, purposeful, aligned with the learning task, and supported by interaction, feedback,

and appropriate instructional design. The distinction is consistent with blended-learning theory (Garrison & Vaughan, 2013), online-learning evidence (Means et al., 2013), learning-analytics studies (Ifenthaler & Yau, 2020; Paulsen & Lindsay, 2024), and digital learning innovation research showing that technology is most valuable when embedded in pedagogical patterns such as collaboration, multimedia, gaming, virtual environments, and blended or flipped designs (Li et al., 2023).

Table 1. Theoretical strands synthesized into the proposed educational-technology model.

Frame work / evidence strand	Core constructs	Main contribution	Remaining issue
Technology acceptance / adoption	Perceived value, effort, social influence, facilitating conditions	Explains why users adopt	Often stops before learning quality
E-learning successes	System, information, service, educational, learner, instructor, support quality	Explains system success and benefits	May not fully model engagement dimensions
Digital engagement	Behavioral, cognitive, affective and social engagement	Explains how learners participate	Can under-specify readiness and institutional context
Learning analytics	Digital traces, feedback, prediction, interventions	Connects data to support	Behavioral indicators can dominate
Proposed model	Readiness -> effective use -> multidimensional engagement	Integrate learner	Requires empirical validation

2.5 Integrated Theoretical Proposition

The theoretical proposition developed in this paper is that student readiness functions as an enabling condition rather than a direct substitute for good pedagogy. Readiness should increase the probability that learners can engage with technology effectively, but the effect on engagement should be stronger when technology is intentionally aligned with instructional

activities and institutional support. In this view, effective use is the mechanism linking readiness to engagement, while facilitating conditions can amplify or constrain the pathway. This proposition reconciles learner-centered evidence with system- and organization-level evidence and provides a coherent basis for the hypotheses presented later.

3. Materials and Methods

3.1 Materials

The evidence base was constructed as a structured integrative review of 26 scholarly sources published from 2010 through 2025. The corpus was intentionally broad enough to cover the major components of the proposed model while remaining focused on higher education and educational technology. Sources included systematic reviews, meta-analyses, empirical studies, conceptual works, and one USJNITE article directly examining technology-enhanced learning, online platforms, engagement, and digital equity (Haider, 2021). Publisher records and open scholarly indexes were used to verify titles, authors, dates, journal information, and DOI details where available. The source mix included work from Springer Nature, Elsevier, PLOS, Frontiers, and established educational-technology venues.

3.2 Search and Eligibility Strategy

The source identification strategy combined searches for educational-technology, digital-learning, readiness, adoption, and engagement concepts. Representative terms included “educational technology,” “digital learning,” “online learning,” “blended learning,” “digital literacy,” “technology self-efficacy,” “technology adoption,” “student engagement,” “learning analytics,” “learning management system,” and “higher education.” Searches were supplemented by backward citation checking from major systematic reviews. A study was eligible when it (a) concerned technology-mediated learning or technology adoption in education, (b) included higher education or findings transferable to higher education, (c) reported substantive empirical, review, or theoretical evidence, and (d) directly informed at least one model component. Commentaries without substantive evidence, purely technical computing papers unrelated to learning, and studies focused exclusively on school-age learners were excluded.

Table 2. Eligibility criteria used to construct the focused review corpus.

Criterion	Inclusion	Exclusion
Publication period	2010-2025	Outside the review window

Criterion	Inclusion	Exclusion
Context	Higher education / tertiary / technology-enhanced learning	Unrelated technology context
Evidence type	Empirical, systematic review, meta-analysis, conceptual framework	Opinion-only commentary
Main constructs	Readiness, adoption, engagement, digital learning, analytics, or related mechanisms	No link to learning/technology outcomes
Accessibility	Verifiable publisher or scholarly bibliographic record	Unverifiable metadata

3.3 Coding Framework

Each study was coded against five analytical layers: learner readiness, technology and system quality, institutional or facilitating conditions, effective technology use, and student engagement. Readiness codes covered digital competence/literacy, technology self-efficacy, technology attitudes/perceived usefulness, and learning readiness. Effective-use codes included usability, quality, interaction, feedback, analytics, and pedagogical alignment. Engagement codes followed a multidimensional structure spanning behavioral, cognitive, emotional, and social/collaborative participation. A study could carry multiple codes; therefore thematic counts are overlapping rather than mutually exclusive.

Table 3. Coding framework used to integrate the literature into the proposed model.

Code family	Operational meaning	Illustrative sources
Digital competence / literacy	Ability to access, understand, evaluate, and use digital tools for learning	Getenet et al.; Bond et al.; Haider
Technology self-efficacy	Confidence in using digital systems to complete learning tasks	Getenet et al.; Al-Fraihat et al.
Technology attitude / usefulness	Positive orientation, expected benefit, satisfaction, or perceived utility	Feng et al.; Al-Fraihat et al.
Facilitating conditions	Infrastructure, support, instructor preparation, access, and policy	Feng et al.; Alqahtani & Rajkhan; Bond et al.
Effective EdTech use	Usability, pedagogical alignment, feedback, analytics, interaction, and purposeful use	Martin et al.; Paulsen & Lindsay; Susnjak et al.

Code family	Operational meaning	Illustrative sources
Student engagement	Behavioral, cognitive, emotional, and social/collaborative participation	Bond et al.; Nkomo et al.; Bergdahl et al.

3.4 Synthesis Procedure

The synthesis used descriptive bibliographic profiling and thematic comparison rather than statistical meta-analysis because the corpus combined heterogeneous designs, constructs, and outcomes. First, publication years were tabulated to show the temporal profile of the evidence base. Second, thematic codes were assigned to the 26 studies. Third, findings were compared across readiness, technology-use, and engagement layers. Finally, recurring relationships were translated into a conceptual pathway: readiness -> effective technology use -> student engagement, with facilitating conditions and equity treated as contextual boundary conditions. The resulting model was checked for conceptual coverage against the major systematic reviews included in the corpus, especially those addressing educational-technology adoption, engagement, learning analytics, and online learning.

4. Results and Discussion

The synthesis produced four broad findings. First, digital learning readiness is multidimensional and cannot be reduced to access or technical skill. Second, educational-technology adoption is consistently influenced by perceived value, usability, quality, and support. Third, student engagement is best treated as multidimensional rather than as a proxy for platform activity. Fourth, the strongest pathway suggested by the evidence is indirect: readiness supports effective use, and effective use creates the conditions for sustained engagement. These patterns are presented below and then integrated into the proposed model.

4.1 Publication Profile of the Review Corpus

The 26-source corpus spans 2010-2025, with a visible concentration after 2020. This pattern reflects both the growth of digital-learning research and the acceleration of educational technology research during and after the COVID-19 period. Reviews of the pandemic literature show that higher education moved rapidly toward remote delivery, producing a large cross-sectional evidence base that focused on student and teacher perceptions (Bond et al., 2021). The earlier literature, by contrast, focused more on the effectiveness of online and blended learning, technology quality, and the design of digitally mediated learning environments (Means et al., 2013; Garrison & Vaughan, 2013; Martin et al., 2020).

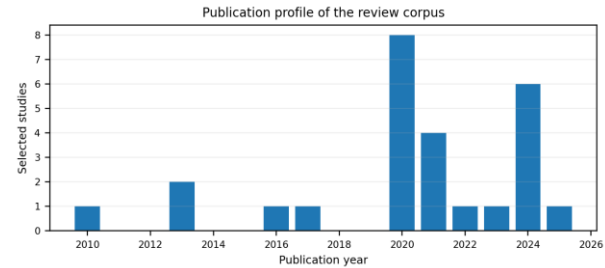


Figure 1. Publication profile of the 26-study review corpus by year.

4.2 Digital Learning Readiness as a Multidimensional Construct

The evidence indicates that readiness has at least four recurring dimensions. The first is digital competence: students need the practical and evaluative ability to navigate platforms, manage digital information, communicate online, and select appropriate tools. The second is technology self-efficacy, or confidence that one can use the technology to complete a learning task. The third is technology attitude and perceived usefulness, which captures whether students believe digital tools are worthwhile and manageable. The fourth is facilitating conditions, including connectivity, device access, technical assistance, instructor support, and institutional policy. Feng et al. (2025) reached a similar conclusion from a systematic review of 39 studies, organizing educational-technology adoption factors into performance expectancy, effort expectancy, social influence, and facilitating conditions.

Getenet et al. (2024) provide especially direct empirical support for the readiness mechanism. In their study of 110 first-year students, positive attitudes toward digital technologies were associated with digital-technology self-efficacy, and self-efficacy was positively related to social, cognitive, behavioral, collaborative, and emotional engagement. The study is important for the proposed model because it demonstrates that the learner-level pathway is not simply “technology -> engagement.” Instead, how students feel about and experience technology influences their confidence, which then shapes engagement.

At the system level, Al-Fraihat et al. (2020) found that successful e-learning depends on a broader collection of quality factors covering system, information, service, educational, support, learner, and instructor dimensions. Their model reinforces the argument that digital readiness should include contextual conditions rather than being treated solely as an individual student characteristic. Similarly, Alqahtani and Rajkhan (2020) emphasized technological, pedagogical, managerial, and learner-related success factors during the rapid expansion of e-learning.

Together, these studies suggest that readiness is partly individual and partly institutional.

From Effective EdTech Use to Multidimensional Student Engagement

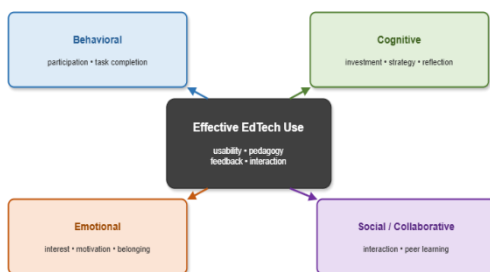


Figure 2. Relationship between effective educational-technology use and multidimensional student engagement.

The engagement pathway in Figure 2 visualizes a measurement principle that recurs across the reviewed literature: technology use should be evaluated by the learning behaviors and experiences it enables, not by activity counts alone. Behavioral participation is only one dimension; cognitive investment, emotional involvement, and social or collaborative interaction provide additional evidence of meaningful engagement. This framing is consistent with the engagement synthesis reported by Bond et al. (2020) and Nkomo et al. (2021), while the learning-analytics literature cautions against relying too heavily on observable digital traces (Bergdahl et al., 2024).

4.3 Student Engagement Is More Than Platform Activity

The literature strongly supports a multidimensional approach to engagement. Bond et al. (2020) mapped 243 studies and reported relationships between digital technologies and behavioral, affective, and cognitive engagement, but also found that few studies clearly defined engagement and fewer than half were guided by a theoretical framework. Nkomo et al. (2021) similarly concluded that digital engagement research spans diverse technologies and constructs and requires attention to the pedagogical appropriateness of the technology rather than the presence of technology alone.

Learning-analytics research adds a measurement warning. Bergdahl et al. (2024) reviewed 159 higher-education studies and found that engagement was overwhelmingly represented through observable behavioral measures, such as clicks and task duration, with relatively little attention to emotional, cognitive, or social dimensions. This is a critical implication for the proposed model: increased platform activity should not automatically be interpreted as deeper engagement. A student can generate many clicks while

experiencing low cognitive investment, or can participate deeply through a smaller number of well-designed learning activities.

The same principle applies to dashboards. Ifenthaler and Yau (2020) found evidence that learning analytics can support study success when analytics are connected to feedback and interventions. Paulsen and Lindsay (2024) observed a movement toward student-facing dashboards that focus more explicitly on learning rather than analytics alone, while Susnjak et al. (2022) emphasized actionable insight, interpretability, and meaningful recommendations. These studies point toward a design principle: data should be translated into feedback and learning actions that students can understand and use.

4.4 Thematic Distribution of the Evidence

The 26 studies were coded across overlapping themes. Engagement appeared in 19 studies, digital literacy or competence in 18, adoption/usefulness in 17, online or blended learning in 16, readiness/self-efficacy in 12, learning analytics in 10, equity/access in 7, and gamification in 5. These numbers describe the emphasis of the selected evidence base rather than the prevalence of themes in the full global literature. The pattern nevertheless shows that engagement and readiness are repeatedly discussed together, while analytics and gamification tend to appear as specific technology mechanisms within a wider learning ecosystem.

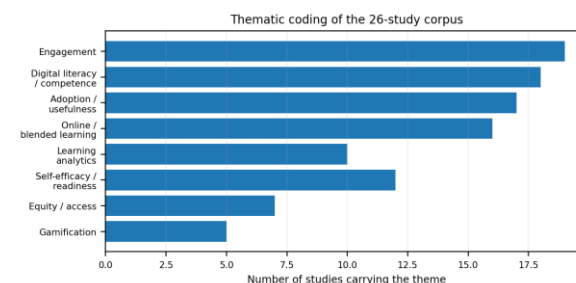


Figure 3. Overlapping thematic coding of the 26-study review corpus.

4.5 Technology Quality, Pedagogy, and Effective Use

One of the clearest results across the literature is that adoption does not guarantee effective use. Martin et al. (2020), in a review of 619 online-learning studies, identified learner characteristics and engagement as frequently examined themes but also highlighted the need for more attention to organizational topics such as leadership, policy, access, culture, equity, inclusion, and ethics. Means et al. (2013) found overall positive effects for online and blended learning relative to face-to-face conditions in their meta-analysis, but effects varied by instructional design and learning conditions. Garrison and Vaughan (2013) similarly frame blended

learning as a deliberate integration of face-to-face and online learning rather than as a technical substitution.

This distinction is central to the present model. Effective educational-technology adoption should be defined as purposeful, usable, pedagogically aligned use that helps learners carry out meaningful activities, not merely log in or submit assignments. Al-Fraihat et al. (2020) provide a quality-based foundation; Martin et al. (2020) add the importance of course and instructor characteristics; and learning-analytics research highlights feedback and intervention. These strands converge on a common idea: the technology becomes educationally valuable when the system, course design, and learner behavior are aligned.

4.6 Engagement-Enhancing Technologies: Analytics and Gamification

The evidence also shows that individual technologies can improve particular engagement mechanisms without becoming the entire explanation of engagement. Learning analytics can identify patterns of activity and provide actionable support. Ifenthaler and Yau (2020) reported evidence that analytics can support study success, but also noted that institutional and technological support conditions matter. Paulsen and Lindsay (2024) and Susnjak et al. (2022) further stress that dashboards should provide interpretable information that students can use to regulate learning.

Gamification presents a complementary case. Sailer and Homner (2020) found small but significant effects across cognitive, motivational, and behavioral outcomes, while Li et al. (2024) found a small overall effect on intrinsic motivation and stronger effects on perceived autonomy and relatedness, with limited effects on competence. Khaldi et al. (2023) found points, badges, leaderboards, levels, feedback, challenges, and storytelling among the most common design elements and noted that personalization is increasingly prominent. These findings support the model's position that a technology feature can act as a mechanism within effective use, but its impact depends on how it is integrated into learning activities and learner motivation.

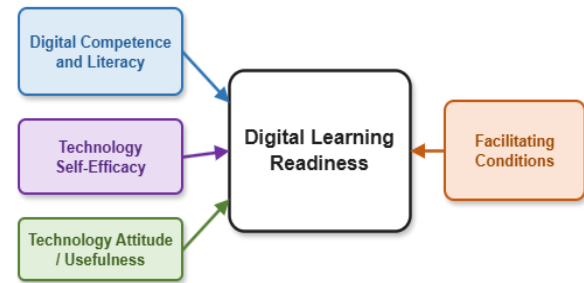


Figure 4. Multidimensional structure of digital learning readiness and contextual enabling conditions.

Figure 4 expands the readiness construct used in the proposed model. The learner-facing dimensions describe what students bring to technology-enhanced learning, while facilitating conditions represent the surrounding infrastructure, support, instructor preparation, and institutional environment. This distinction prevents readiness from being interpreted as an individual deficit and reflects the review evidence that technology success depends on both learner characteristics and contextual conditions (Al-Fraihat et al., 2020; Feng et al., 2025; Warschauer & Matuchniak, 2010).

4.7 Equity, Access, and the Risk of Treating Readiness as an Individual Deficit

A readiness model that focuses only on students risks blaming learners for conditions they cannot control. Warschauer and Matuchniak (2010) distinguish between access to technology and differences in skill and quality of use, while Haider (2021) reports that digital equity and access remained central challenges in the technology-enhanced learning literature leading into and through the pandemic period. Bond et al. (2021) also showed that much of the emergency remote teaching evidence was cross-sectional and perception-oriented, indicating the importance of studying long-term institutional and pedagogical conditions.

For this reason, facilitating conditions are positioned in the model as a boundary condition rather than a minor predictor. A student with strong digital literacy may still experience low engagement when bandwidth is unstable, a platform is poorly designed, support is unavailable, or the instructor uses technology mainly for content transmission. Conversely, a strong support environment can reduce the friction between readiness and participation. This interpretation also aligns with Feng et al. (2025), who identify facilitating conditions as one of the four core adoption dimensions.

4.8 Proposed Model for Effective Educational Technology Adoption

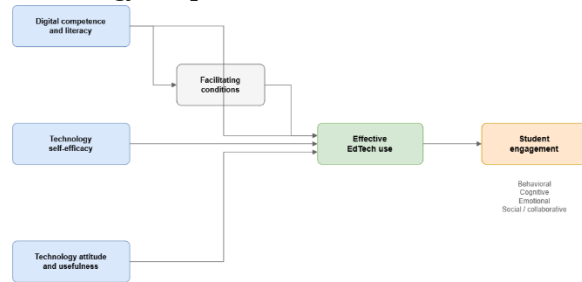


Figure 5. Integrated model linking digital learning readiness, effective educational-technology use, and student engagement.

The proposed model contains three levels. At the first level, learner readiness is represented by digital competence/literacy, technology self-efficacy, and technology attitude/perceived usefulness. At the same level, facilitating conditions capture infrastructure, support, instructor readiness, and institutional policies. The second level is effective educational-technology use, defined as purposeful and pedagogically aligned use characterized by usability, meaningful interaction, timely feedback, learning analytics where appropriate, and suitable instructional design. The third level is student engagement, represented through behavioral, cognitive, emotional, and social/collaborative dimensions.

The model predicts that readiness factors increase the likelihood of effective technology use, but that readiness will not translate into engagement automatically. Effective use is the proximal mechanism: when students can use technology confidently, see its value, and operate in supportive conditions, technology is more likely to facilitate useful learning activity. Engagement then emerges from the interaction among learner readiness, course design, instructor practice, and technology features. This pathway is consistent with Getenet et al. (2024), Al-Fraihat et al. (2020), Martin et al. (2020), Bond et al. (2020), and the more recent adoption review by Feng et al. (2025).

Table 4. Proposed relationships for future empirical validation of the model.

Proposed hypothesis	Relationship	Rationale from the synthesis
H1	Digital competence/literacy -> Effective EdTech use	Students with stronger digital competence are expected to use technologies

Proposed hypothesis	Relationship	Rationale from the synthesis
H2	Technology self-efficacy -> Effective EdTech use	Confidence should reduce technology-related friction and support independent learning behavior.
H3	Technology attitude/usefulness -> Effective EdTech use	Positive value perceptions should increase willingness to integrate technology into learning tasks.
H4	Facilitating conditions -> Effective EdTech use	Infrastructure, support, and instructor preparation should enable sustained and usable adoption.
H5	Effective EdTech use -> Student engagement	Purposeful, interactive, and pedagogically aligned use should strengthen engagement.
H6	Facilitating conditions moderate readiness -> use	Strong contextual support should strengthen the translation of readiness into effective use.
H7	Effective EdTech use -> multidimensional engagement	The strongest benefits should extend beyond behavioral activity to cognitive, emotional

4.9 Comparison with Prior Reviews

The proposed synthesis differs from single-topic reviews because it treats readiness, technology quality, and engagement as connected layers. Earlier evidence maps such as Bond et al. (2020) established the breadth of research linking digital technology and student engagement, but also highlighted weak theoretical framing and concentration in a limited set

of countries. Martin et al. (2020) organized online-learning research by learner, course/instructor, and organizational levels and identified a need for more organizational-level evidence. Ifenthaler and Yau (2020) focused on learning analytics and study success, showing that analytics can support continuation and completion when used with meaningful interventions. More recent reviews have advanced the field by examining engagement measurement in learning analytics (Bergdahl et al., 2024), dashboard design (Paulsen & Lindsay, 2024), and educational-technology adoption factors (Feng et al., 2025).

The gap left between these reviews is an integrative learner pathway. Adoption research explains why technologies are accepted; learning-quality research explains which system and course characteristics matter; engagement research explains how students participate; and analytics research explains how digital traces can inform support. The present model connects these strands by positioning digital readiness as the capacity to use technology, effective use as the mechanism that converts capacity into learning activity, and engagement as the multidimensional outcome. This framing also makes it possible to test where the pathway weakens: for example, a student may have high readiness but poor course design, or a well-designed course may still fail when access and institutional support are weak.

5. Discussion

The synthesis suggests that the central problem in educational-technology adoption is not adoption itself but the conversion of technological capability into learning value. This conclusion is consistent with several strands of evidence. Adoption reviews emphasize usefulness, effort, social influence, and facilitating conditions; quality research emphasizes system, information, service, pedagogical, learner, and instructor factors; engagement research emphasizes multidimensional participation; and analytics research emphasizes the need to connect data to learning actions. The proposed model integrates these strands by putting the learner in the center while explicitly retaining the institutional environment.

First, readiness should be understood as a state of preparedness for learning with technology rather than as a checklist of software skills. A student may know how to operate a learning-management system yet remain digitally unprepared for information evaluation, collaborative work, self-regulation, or sustained online participation. The evidence from Getenet et al. (2024) is particularly useful here because it links attitudes and literacy with digital self-efficacy and then connects self-efficacy with multiple

engagement dimensions. This suggests that readiness interventions should combine technical competence with confidence-building and authentic learning practice.

Second, effective adoption requires pedagogical alignment. Means et al. (2013) showed that online and blended learning can generate positive outcomes, but technology does not function as an independent treatment. Garrison and Vaughan (2013) likewise frame blended learning as an intentional pedagogical design. Al-Fraihat et al. (2020) show that system quality, educational quality, support, learner and instructor factors operate together. Thus, a university that measures adoption only through login counts or platform penetration may overestimate educational value.

Third, engagement should be measured with greater conceptual discipline. Bond et al. (2020) found that relatively few studies explicitly defined engagement, while Bergdahl et al. (2024) identified strong reliance on behavioral log data. For institutional decision-making, this means a dashboard should not equate activity with learning. A mature engagement framework should combine behavioral indicators with student perceptions, cognitive investment, collaborative participation, and emotional experience. This is especially important when analytics are used to trigger interventions.

Fourth, analytics and gamification should be treated as design mechanisms within the larger model rather than as universal solutions. Learning analytics can support study success and help identify students who may need support, but ethical data use, interpretability, and intervention design remain important. Gamification can improve motivation and some learning outcomes, but effect sizes vary and some design approaches have limited impact on competence. These findings reinforce the broader model: technology features work when they are aligned with learner needs, course design, and support structures.

Finally, equity must remain a central concern. The digital divide is not only about who has a device. Differences in skills, quality of use, connectivity, accessibility, and institutional support shape the experience of technology-mediated learning. The model therefore rejects the assumption that low engagement is always a student readiness problem. Researchers and institutions should distinguish learner-level readiness from structural barriers and report both. This would also help address the geographic concentration and limited Global South coverage identified in major engagement evidence maps (Bond et al., 2020).



Figure 6. Six-stage institutional implementation roadmap for effective educational-technology adoption.

The roadmap in Figure 6 converts the conceptual model into an implementation sequence. Diagnosis precedes design, and design precedes wider deployment so that institutions can identify readiness barriers before scaling a platform. The pilot and monitoring stages are deliberately iterative: evidence from student experience, analytics, and course-level outcomes can be used to refine the design before institutional expansion. This sequence reinforces the paper’s central argument that effective adoption is a pedagogical and organizational process, not simply a technology procurement decision (Garrison & Vaughan, 2013; Martin et al., 2020).

5.1 Practical Implications

- Universities should assess digital readiness before introducing new platforms, but the assessment should combine competence, self-efficacy, attitudes, access, and support conditions rather than a single digital-skills score.
- Course designers should evaluate whether technology changes the learning activity for the better. The goal is not maximum tool use; it is better interaction, feedback, collaboration, self-regulation, and access to useful learning resources.
- Institutions implementing learning analytics should monitor multiple engagement dimensions and explain analytics outputs to students. Dashboards should support action rather than merely display data.
- Professional development should focus on pedagogical redesign as well as technical operation. Emergency-remote teaching experience should not be treated as a direct proxy for intentionally designed online or blended learning (Hodges et al., 2020; Bond et al., 2021).
- Equity policies should include connectivity, accessibility, device support, digital-skills development, and responsive technical assistance so that readiness can be developed rather than assumed.

5.2 Institutional Implementation Roadmap

The model can be translated into an institutional implementation sequence. The first stage is diagnosis: universities should establish a baseline profile of student competence, self-efficacy, attitudes, access, and support needs. The second stage is design:

technology should be mapped to course outcomes and learning activities before the platform is configured. The third stage is enablement: students and instructors need orientation, technical support, accessibility arrangements, and opportunities to practice with the technology in authentic tasks. The fourth stage is piloting, followed by mixed-method monitoring and iterative scaling. This approach treats adoption as an ongoing educational process rather than a procurement event.

Table 5. Illustrative institutional roadmap derived from the integrated model.

Stage	Key action	Expected output
1. Diagnose	Readiness, access, support, and course needs	Baseline readiness profile
2. Design	Align tools with outcomes, activities, interaction, and feedback	Pedagogically aligned learning design
3. Enable	Training, accessibility, orientation, technical support	Reduced implementation friction
4. Pilot	Trial in selected courses and collect mixed evidence	Context-specific evidence
5. Monitor	Combine analytics with survey/interview evidence	Actionable engagement signals
6. Scale	Refine policy, infrastructure, and professional learning	Sustainable adoption

The roadmap also reflects an important implication of the evidence on digital equity. A readiness diagnosis should not be used to label students as “ready” or “unready.” Instead, it should identify which conditions can be strengthened. For example, a low self-efficacy score may call for orientation and guided practice, while a connectivity problem requires an institutional infrastructure response. Similarly, low engagement may reflect course design or assessment problems rather than a lack of student motivation. This distinction can help institutions use readiness data to improve environments instead of merely screening learners.

Future Empirical Validation Pathway



Figure 7. Proposed empirical validation pathway for the integrated educational-technology model.

Figure 7 organizes the future research agenda into progressively stronger forms of evidence. Initial work should establish reliable measures for readiness, effective use, and engagement; subsequent multi-institutional studies can test the proposed relationships, while longitudinal designs can examine whether readiness and engagement remain stable over time. Mixed-method and experimental studies can then test how particular technology mechanisms, such as analytics, feedback, gamification, or personalization, contribute to learning under different contexts. The sequence directly addresses the limitations identified in the reviewed literature, particularly cross-sectional designs, narrow behavioral measures, and limited evidence on long-term effects (Bergdahl et al., 2024; Ifenthaler & Yau, 2020; Paulsen & Lindsay, 2024).

6. Future Directions

- Empirical validation of the proposed model using structural equation modelling across multiple universities and disciplines.
- Longitudinal studies that test whether increases in readiness produce sustained engagement over more than one academic term.
- Cross-country comparisons that explicitly examine differences in infrastructure, digital equity, language, and institutional support.
- Multi-method engagement measurement that combines surveys, learning analytics, interviews, and authentic learning-performance indicators.
- Experimental research on which combinations of analytics, feedback, gamification, and personalized

learning features are most effective for different learner profiles.

These priorities respond directly to the limitations repeatedly identified in the literature: heavy reliance on cross-sectional perception data, concentration on behavioral engagement indicators, uneven theoretical grounding, and insufficient attention to organizational and equity dimensions (Martin et al., 2020; Bond et al., 2020; Bergdahl et al., 2024).

7. Limitations

This demonstration paper uses a focused 26-source review corpus rather than a fully exhaustive database search covering every publication in the global educational-technology literature. The purpose is to build and illustrate a coherent model for later empirical testing.

The studies vary substantially in design, population, setting, and technology type. Consequently, thematic counts are descriptive and overlapping and should not be interpreted as effect sizes or as estimates of field-wide prevalence.

Several foundational sources predate current post-pandemic developments. Although the corpus includes recent 2024-2025 evidence, additional emerging work may change the relative importance of particular technologies and constructs.

The proposed hypotheses are derived from synthesis and are not presented as statistically tested findings. A future primary study should validate the measurement model, test common-method bias, and examine alternative pathways and moderators.

8. Conclusion

This paper developed an integrated model of digital learning readiness and student engagement for effective educational-technology adoption in higher education. The literature shows that readiness is multidimensional: digital competence, technology self-efficacy, technology attitude/perceived usefulness, and facilitating conditions repeatedly appear as important components. The evidence also shows that engagement cannot be reduced to platform activity; behavioral, cognitive, emotional, and social dimensions should be considered together.

The resulting model proposes that readiness creates the capacity for technology use, while effective, pedagogically aligned use acts as the mechanism through which technology can support engagement. Institutional support and equity shape the strength of this pathway. This framing moves the discussion beyond technology adoption as an installation problem and toward educational technology as a learning-design and learner-development problem.

The principal implication is straightforward: universities should not ask only whether students are using educational technology. They should ask whether students are ready to use it, whether the learning design makes that use meaningful, whether institutional conditions support it, and whether engagement improves in ways that matter for learning. The proposed model provides a testable structure for answering these questions and offers a basis for future empirical research on sustainable, student-centered educational technology adoption.

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