# Incorporating face-to-face and online learning features to propose blended learning framework for Post-COVID classrooms in India

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## Abstract

**Purpose** – This study examines learner readiness and satisfaction with blended learning systems in India's post-vaccine classrooms, focusing on the relationship between face-to-face (F2F), online learning (OL), and blended learning (BL) indicators and identifying which predictors within these systems most significantly affect learners' satisfaction (LS).

**Design/methodology/approach** – An online survey was conducted with 451 students from both public and private universities in India. The data were analyzed using factor analysis and confirmatory factor analysis, followed by multiple regression to test the hypotheses.

**Findings** – The findings reveal a significant positive correlation between learners' satisfaction and online learning and blended learning, with  $\beta$  values of 28.3 and 27.2, bearing a *p*-value of 0.000. In contrast, face-to-face (F2F) learning was insignificant, with a  $\beta$  value of 0.070 and a *p*-value of 0.119. These insights underscore the effectiveness of online and blended learning formats in enhancing learner satisfaction in higher education while also suggesting a re-evaluation of the role of traditional F2F learning methods. The research supports the integration of online learning in higher education due to its balanced mix of teacher-led and student-centered instruction, alongside the practical benefits of reduced travel costs and access to independent study resources.

**Practical implications** – This study provides insights into student perceptions and attitudes towards blended learning in India's post-vaccine classrooms. It highlights the importance of tailoring blended learning strategies to meet colleges' and universities' diverse learning needs and goals in this evolving context. The findings serve as a valuable resource for educators and administrators, aiding in designing effective blended learning frameworks suited explicitly for higher education in India.

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**Originality/value** – While there is extensive literature on F2F, OL, and BL, limited research compares these learning approaches and their impact on learner satisfaction within the Indian higher education context. This study fills this gap by providing valuable insights for shaping educational strategies in India's post-vaccine classrooms. **Keywords** Blended learning, Face-to-face learning, Learners' perceptions, Online learning, Post-COVID-19 class

Paper type Research paper

## 1. Introduction

Throughout history, the highly infectious coronavirus has significantly impacted human life, confining us indoors, causing millions of deaths, and halting many aspects of life. including global education (Singh et al., 2021; Aristovnik et al., 2023; Tang, 2023; Dedeilia et al. 2023). The education sector suffered immensely due to skyrocketing coronavirus cases and faced numerous challenges in teaching and learning, which necessitated a rapid transformation in educational methodologies. This situation increased the workload for teachers and faculty at institutions like schools, colleges, and universities (Dhawan, 2020; Rapanta et al., 2020). The rising death toll and ongoing spread of the virus raised concerns among students, teachers, and educators regarding the timely reopening of educational institutions, further extending the reliance on online education. Previous works have documented the progression of coronavirus from an epidemic to a pandemic, suggesting that COVID-19 would not quickly disappear and emphasizing the need for persistent COVID-19-safe practices to prevent its spread (Singh et al., 2021). This scenario has intensified fear and anxiety about the future of in-person learning worldwide. Nonetheless, there has been promising news from the scientific community regarding the development of vaccines to boost the human body's defence against the virus. Studies indicate that vaccines provide essential antibodies to disrupt the virus's transmission chain. However, even with vaccination, it is strongly recommended that students continue to exercise caution and adhere to COVID-19 safety guidelines in educational settings (Powell et al., 2021). In these circumstances, a blended or hybrid learning model, combining in-person and online learning, emerges as a viable solution to meet educational needs amid these precautions and restrictions.

Teaching-learning practice is rapidly changing due to the consecutive waves of COVID-19. Institutes are quickly reshaping their pedagogical methods and moving towards blended learning (BL) and remote learning environments if some have not done the same (Ehrlich *et al.*, 2020; Moszkowicz *et al.*, 2020). The BL is a pedagogical method that combines online and traditional or face-to-face learning opportunities (Garrison and Kanuka, 2004; Garrison and Vaughan, 2008). This technique offers several key features: (1) learner-centric instruction, where learning contents are designed according to the student's individual preference; (2) focus given on the enhancement of teacher-learner, learner-learner, learner-educational contents, and learner-other stakeholders interactions, including learner-supplementary materials; and (3) continuous collection of formative and summative testing results to upgrade the courses (Watson, 2008).

Due to the continuous COVID-19 waves, BL is being rapidly adopted as a pedagogical method globally. Nevertheless, it is an overnight change and hassle-free. Though BL is an old practice in developed nations, developing countries still face numerous challenges regarding its smooth and effective application. Because BL's effectiveness greatly depends on several factors, one such problem is ensuring the hassle-free accessibility of technology to learners (Hofmann, 2014). He further mentioned that the difficulty in technology accessibility might lead to both users' abundance of learning and the failure of the technical learning applications. According to a report, 16% of learners have negative attitudes towards blended learning, whereas 26% believe the course won't be completed (The Oxford Group, 2013).

Learners are the key participants in teaching-learning activities. As a result, it is vital to Asian Association determine their readiness and perceptions for effective pedagogical implementation that fits well with their backgrounds and abilities, especially in a BL context, as some learners may not cope with the new design and tools. Hence, their academic performance may be negatively impacted. Therefore, the existing study examines university learners' perceptions of adopting BL, which is also likely to happen in post-COVID classrooms in India. In this regard, students' demographic profiles have been inquired to assess learners' attitudes towards it in post-COVID classes concerning the prior experience of BL, required technical skills, use of technology for teaching-learning activities, education level, age, gender, etc. Furthermore, we have also investigated the content quality, content design, and features such as flexibility and anytime/anywhere, including their role in goal achievement that attracts learners towards BL. Furthermore, overall attitudes towards traditional teaching (F2F learning), pure online learning (OL), and blended learning and their overall impact on learners' academic satisfaction.

## 2. Theoretical underpinnings and study framework

COVID-19 profoundly impacted academia, presenting unprecedented challenges during the 2020–2021 academic session. Teachers, learners, and educational administrators faced significant barriers due to the pandemic (Singh et al., 2021). Despite the widespread administration of vaccines, transitioning back to traditional in-person classrooms was fraught with challenges, influenced by the ongoing impact of the coronavirus (Dorn et al., 2020).

Before the pandemic, the educational landscape predominantly operated in a traditional or pure face-to-face (F2F) format. This approach offered several benefits, such as direct interaction between instructors and learners, fostering lively in-class discussions, and enabling immediate question-and-answer sessions (Paul and Jefferson, 2019). However, those accustomed to F2F learning found online learning challenging, as it often required extended periods of computer use (Roval and Jordan, 2004; Qamar et al., 2023). Numerous studies have highlighted that F2F learning can significantly benefit students, from enhancing motivation and fostering community-building bonds to enabling teachers to make decisions and choose the right content and instructional methods (Paul and Jefferson, 2019).

Despite the advantages of F2F learning, the pandemic necessitated a swift shift to online learning for schools, colleges, and universities to maintain uninterrupted education (Singh et al., 2021; Singh and Matthees, 2021). Online learning provides benefits like self-regulated learning, improved time management, collaborative learning, flexibility, and costeffectiveness (Singh and Matthees, 2021; Smith and Hardaker, 2000). The persistent surge in COVID-19 cases meant that online education remained a primary learning mode. With the advent of vaccinations, the situation gradually started to relax, leading to the adoption of flipped, hybrid, or blended learning models. These models enabled learners to participate in education with appropriate COVID-19 safety measures, as health experts predicted that more virus waves would potentially restrict outdoor activities (Singh and Matthees, 2021).

The pandemic-induced transition in the education sector has effectively transformed the higher education structure. Every learning format, whether traditional, online, or blended, has advantages and disadvantages. In the post-pandemic scenario, numerous educational institutions have transitioned from a purely online format to a blended learning approach. Blended learning combines traditional F2F classroom sessions with online learning sessions, offering students the advantages of both formats. This includes flexible schedules, self-paced homework, and assignments (Singh et al., 2021). This instructional approach is a paradigm shift, allowing educators to redesign and update educational content to suit students' specific needs, especially in courses requiring more engaging learner experiences, which is challenging to achieve in purely F2F or online formats (Dhawan, 2020).

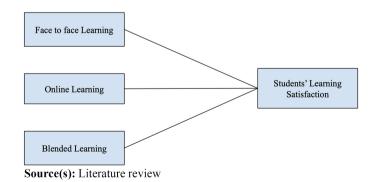
Our study aims to identify key indicators for effectively implementing blended learning. These indicators are crucial for understanding how blended learning affects learner satisfaction during pandemics and how educators can utilize these indicators while planning post-pandemic teaching strategies (see Figure 1). While there is a rich body of literature on the effectiveness of F2F and online learning, there is a notable gap in studies focusing on blended learning effectiveness indicators. These indicators are essential for providing a meaningful learning environment in higher education. Systematic investigation of these indicators can significantly benefit the teaching-learning community, especially with the resumption of inperson education and the possibility of future COVID-19 waves. Such research will assist educational institutes, app developers, and other stakeholders in providing continuous education through blended learning while adhering to COVID-19-appropriate behavior.

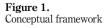
Therefore, we propose a state-of-the-art model based on our study's key findings. This model outlines implementing blended learning effectively in higher education settings (see Figure 5). The existing literature shows only a few studies focusing on blended learning effectiveness indicators, which instructors can leverage to provide high-quality education to learners. Our research seeks to fill this gap by offering a comprehensive view of these indicators, explicitly based on quantitative and qualitative analyses. At the same time, extensive literature covers the advantages and disadvantages of both F2F and online learning formats. It is pertinent to systematically investigate blended learning key effectiveness indicators that could facilitate instructors in providing a meaningful learning environment, especially in higher education. Such findings will surely benefit teaching-learning communities, particularly as we transition back to in-person education due to the declining trend in COVID-19 cases. Additionally, as health experts predict more waves in the future, our study will help educational institutes, app developers, and other stakeholders provide continued education through blended learning along with appropriate COVID-19 behaviour wherever and whenever necessary.

## 3. Related works and hypothesis development

## 3.1 Face-to-face learning and learners' satisfaction

Over the past 2 decades, classroom scenarios have undergone significant changes due to the integration of technology and its inevitable penetration into human life (Tratnik *et al.*, 2019). With the COVID-19 outbreak, the education system experienced a dramatic shift from offline to online modalities (Lee *et al.*, 2021), leading to exponential growth in various online courses. The nature of online courses differs considerably from face-to-face courses, primarily in that students typically have more freedom to control their learning progress. In contrast, face-to-face learning is generally more teacher-centric, offering students less control over their





learning process (Tratnik et al., 2019; Hamid, 2001). Despite teachers showing significant Asian Association progress in interest and creativity in online courses (see Otter et al., 2013; Weber and Lennon, 2007: Zhang et al. 2010), the literature comparing pure face-to-face and pure online learning courses has recorded mixed learner reactions. Few studies reported higher satisfaction with face-to-face learning (Allen and Seaman, 2004; Roach and Lemasters, 2006; Tallent-Runnels et al., 2006), others preferred online learning (Hui et al., 2008), and some found no statistical difference between the two formats (Sitzmann et al., 2006; Zhang et al., 2004; Tratnik et al., 2019). Consequently, this study compares students' overall satisfaction with all three face-toface, online, and blended learning formats among Indian university learners. Specifically, this paper seeks to identify key effectiveness indicators affecting students' learning satisfaction in relation to education content, delivery of material, course features, the role of the teacher, teacher-student interaction frequency, self-regulation, and student-student interactivity. Therefore, face-to-face learning is considered a critical indicator of students' learning satisfaction with the system, leading to the following hypothesis:

H1. Face-to-face learning positively affects students' learning satisfaction.

#### 3.2 Online learning and learners' satisfaction

Previous studies highlight the critical role of learner satisfaction—shaped by attitudes, system usability, and experiences—in sustaining engagement and determining success in online learning environments, ultimately influencing learning outcomes (Dhawan, 2020; Moore and Kearsley, 2012; Parahoo et al., 2016; Ke and Kwak, 2013).

The pandemic-induced transition to online learning prompted research into educational needs and learner sentiment, identifying eight barriers to satisfaction: technical issues, administrative challenges, inadequate pedagogical and technical skills, interactivity problems, poor connectivity, accessibility issues, low motivation, internet cost concerns, limited availability, and insufficient study support. These factors are critical to understanding and improving the online learning experience (Hew et al., 2020; Jiang et al., 2021: She *et al.*, 2021: Muilenburg and Berge, 2005).

Baber (2020) compared perceptions of online learning system satisfaction indicators between Indian and South Korean undergraduate students. The findings indicated that students' learning satisfaction was influenced by factors such as classroom interactivity, engagement, course flow, instructor skills, and positive teacher attitudes. Additionally, aspects related to system support, information, and the quality of the online platform's services were significant. Students also noted that self-control, advanced digital skills, prior system knowledge, and readiness to use smart tools were crucial for satisfaction in online courses (Lee et al., 2011; Jan, 2015; Jiang et al., 2021).

Thus, the current study recognizes online learning as a key indicator of students' learning satisfaction and hypothesizes that:

H2. Online learning positively affects students' learning satisfaction.

### 3.3 Blended learning and learners' satisfaction

Blended Learning (BL) combines traditional classroom instruction with online learning and is emerging as a popular, flexible instructional strategy. This approach, often seen in flipped classrooms, shifts the focus from teacher-led to learner-driven experiences, allowing students to learn at their own pace through online materials and in-person sessions. BL emphasizes collaborative learning in face-to-face sessions through group activities and problem-solving, thereby enhancing critical thinking and cognitive skills. This evolution from conventional to more interactive and student-focused learning methodologies underscores BL's role in modernizing educational practices (Lin et al., 2022; McCarthy, 2016a, b).

In BL environments, the role of the student is amplified; they have increased opportunities to interact with teachers, peers, and other students. Teachers act as facilitators or collaborators, providing enhanced guidance, mentoring, tutoring, and interaction, contributing to students' increased satisfaction with the course (McCarthy, 2016a, b; Moffett, 2015). Previous studies have identified factors that could hinder the effectiveness of BL learning outcomes and reduce student satisfaction. These factors are primarily categorized into pedagogical and technological components. Issues such as poor quality of educational content design, video production quality, and video resolution have been cited as reasons for student dropout (Akçayır and Akçayır, 2018; McCarthy, 2016a, b; Milman, 2012), as well as the degree of learning from these videos (Milman, 2012).

Findings from previous research on BL have been varied, depending on the constructs used to measure student satisfaction. Some studies have noted enhanced performance and satisfaction with BL courses (Låg and Sæle, 2019; Strelan *et al.*, 2020) compared to pure face-to-face and online learning. However, other students have expressed higher dissatisfaction and lower academic performance in BL courses than in traditional learning settings (Van Alten *et al.*, 2019). In this context, our study aims to compare the components of pure face-to-face, online, and blended learning to uncover nuanced insights into students' perceptions of BL. We hypothesize that:

H3. Blended learning positively affects students' learning satisfaction.

### 4. Method

4.1 Sample, data collection, and questionnaire development

In the evolving landscape of Indian education, particularly in post-vaccine classrooms, the transition from traditional face-to-face (F2F) to online learning (OL) during the initial COVID-19 lockdown has significantly shifted the psychological barriers of educational stakeholders toward technology in teaching. Despite a wealth of literature on F2F and OL, there's a notable gap in studies comparing these methods to develop blended learning (BL) in India. Recognizing this, the University Grants Commission (UGC) has advocated for integrating BL into educational institutions. This study aims to merge factors from F2F, OL, and BL, assessing their impact on student satisfaction to formulate an effective BL framework for India.

The research delves into the existing literature on F2F, OL, and BL (Singh *et al.*, 2021; Gherhes *et al.*, 2021; Bowden, 2022; Atwa *et al.*, 2022; Singh *et al.*, 2022) to develop a comprehensive research instrument. A questionnaire focusing on educational material quality, content delivery, and interactivity was created to measure its impact on student satisfaction. A pilot study with 76 students refined the questionnaire, which two experienced teachers, one in language and another in statistics, further improved. Both experts, familiar with F2F and online courses during the pandemic, rigorously validated the questionnaire, leading to the deletion of four items due to semantic redundancy and the addition of one item related to system ease of use.

The finalized questionnaire, consisting of 28 items, aimed to assess students' learning satisfaction across F2F, OL, and BL, reflecting their experiences through pre-pandemic, pandemic, and post-vaccine phases. It also included demographic queries. A preliminary test with three students ensured the clarity and consistency of the questions.

Subsequently, 560 students from two Indian universities – a private university in Bengaluru and another in New Delhi – participated in the survey. All respondents had experience with all three learning formats. The survey introduction clarified the study's objectives, data usage, and confidentiality. The data collection spanned from January 2021 to August 2022. Responses were ranked on a 5-point Likert scale from complete agreement to

total disagreement. After the data collection, 13 unsuitable responses were excluded, leaving Asian Association 451 for quantitative analysis to identify key determinants of F2F, OL, and BL. of Open

5. Data analysis

## 5.1 Demographics

Tha data for the study were gathered from 451 participants, comprising 281 males and 170 females, all aged between 21 and 38 years. Seventy-six (76) were married, and 375 were unmarried. The educational background revealed 183 graduates. 247 postgraduates, and 21 doctorates. In the following sections, an analysis of the data is presented.

## 5.2 Reliability analysis

The reliability analysis for the face-to-face learning items resulted in the retention of six variables post-deletion in the Exploratory Factor Analysis (EFA), vielding a Cronbach's alpha of 0.732. In contrast, the original compositions of the online and blended learning items were maintained, each consisting of 10 items, with Cronbach's alpha values of 0.860 and 0.832, respectively. This indicates a 73.2%, 86%, and 83.2% consistency under the independent variable category. Moreover, the dependent variable, learning satisfaction, demonstrated high consistency with a Cronbach's alpha of 0.946. (see Table 1).

## 5.3 Exploratory factor analysis – face-to-face, online, and blended learning

Exploratory Factor Analysis (EFA) of variables from literature-based questions identified key factors for face-to-face, online, and blended learning. Four items in the face-to-face category were excluded due to poor fit. EFA outcomes showed a Kaiser-Meyer-Olkin (KMO) measure of 0.632 for face-to-face learning, indicating moderate sampling adequacy, and a Bartlett's test of sphericity was not significant (b > 0.05), suggesting items may not be interrelated as expected. The total variance explained was 61.4% for face-to-face learning. Online and blended learning analyses vielded better KMO values of 0.846 and 0.797. respectively, both with Bartlett's test of Sphericity results also above the 0.05 threshold, and total variances explained by their factors were similarly high at 61.3 and 61.4%. Variables across all learning modes were grouped into two distinct categories under the rotated component matrix, indicating good data quality for further causal relationship assessment toward learner satisfaction. The precise categorization and full statistical details are provided in Table 2 of the study.

## 5.4 Measurement model – face-to-face, online, and blended learning

5.4.1 Measure for the goodness of fit and estimates. In the initial model run, the face-to-face learning model exhibited the following indices: CMIN/DF = 3.04, Chi-square = 65.9,

|                         | No of items | Cronbach's alpha value |                      |
|-------------------------|-------------|------------------------|----------------------|
| Independent variables   |             |                        |                      |
| Face-to-face learning   | 6           | 0.732                  |                      |
| Online learning         | 10          | 0.860                  |                      |
| Blended learning        | 9           | 0.832                  |                      |
| Dependent variables     |             |                        |                      |
| Learning satisfaction   | 12          | 0.946                  | Table 1.             |
| Source(s): Primary data |             |                        | Reliability analysis |

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| AAOUJ   |   |   |                        | VMO  | 1.D   | 4 4                          |  |   |                                  |
|---|---|---|------------------------|--|---|------------------------------|--|---|----------------------------------|
| v   | Face-to- face lea   | rning                                     |                        |  | d Bartlett's<br>ne learning                             | test                         | Blended learning   |   |                                  |
|   | Kaiser-Meyer-Ol<br>Measure of Sam<br>Adequacy                                     |   | 0.632                  | Kaiser-Meyer<br>Measure of S<br>Adequacy   |   | 0.846                        | Kaiser-Meyer<br>Measure of S<br>Adequacy                     |   | 0.797                            |
|   | Bartlett's A<br>Test of   | Approx. 23<br>Chi-<br>Square              | 0.808                  | Bartlett's<br>Test of<br>Sphericity  | Approx.<br>Chi-<br>Square                               | 903.664                      | Bartlett's<br>Test of<br>Sphericity                          | Approx.<br>Chi-<br>Square                 | 709.608                          |
|   | Total variance<br>explained (%)   | Df<br>Sig                                 | 15<br>0.000<br>51.47   | Total variance explained (%  | Df<br>Sig<br>æ  | 45<br>0.000<br>61.31         | Total variand<br>explained (%                                | Df<br>Sig<br>ce                           | 36<br>0.000<br>61.40             |
|   | Face-to- face learning<br>Components  |   |                        | Rotated component matrix <sup>a</sup><br>Online learning<br>Components<br>Indicators 1 2 |   |                              | Blended learning<br>Components<br>Indicators 1 2             |   |                                  |
|   | FTF6<br>FTF8<br>FTF1<br>FTF3<br>FTF5  | 0.874<br>0.826<br>0.710<br>0.<br>0.<br>0. | 2<br>834<br>685<br>601 | OL5<br>OL6<br>OL7<br>OL8<br>OL9<br>OL10<br>OL1<br>OL2<br>OL3                             | 1<br>0.804<br>0.784<br>0.784<br>0.746<br>0.720<br>0.707 | 2<br>0.871<br>0.838<br>0.800 | BL5<br>BL6<br>BL7<br>BL10<br>BL9<br>BL1<br>BL2<br>BL3<br>BL4 | 0.824<br>0.822<br>0.758<br>0.717<br>0.650 | 0.821<br>0.777<br>0.732<br>0.635 |
| <b>Table 2.</b><br>Exploratory factor<br>analysis | Note(s): Extra<br>normalization<br><sup>a</sup> Rotation conve<br>Source(s): Prin | erged in 3 iter                           | -                      | ncipal compo   | nent analy  | sis rotatio                  | on method: v   | arimax with                               | h Kaiser                         |

p-value = 0.000, GFI = 0.939, NFI = 0.863, CFI = 0.875, and RMSEA = 0.041. The online learning model showed CMIN/DF = 3.19, Chi-square = 199.9, p-value = 0.000, GFI = 0.899, NFI = 0.880, CFI = 0.893, and RMSEA = 0.037. The blended learning model reported CMIN/DF = 5.87, Chi-square = 177.8, p-value = 0.000, GFI = 0.919, NFI = 0.911, CFI = 0.922, and RMSEA = 0.053. These results indicated a relatively poor fit in the parsimonious model run.

Subsequently, an improved model run was conducted, where two items were deleted for online and blended learning. The revised fit indices indicated a good model fit for all three independent variables. The face-to-face learning showed improved results with CMIN/ DF = 2.17, Chi-square = 62.7, *p*-value = 0.000, GFI = 0.958, NFI = 0.869, CFI = 0.880, and RMSEA = 0.037. The online learning model also demonstrated a good fit with CMIN/ DF = 2.81, Chi-square = 196.2, *p*-value = 0.000, GFI = 0.919, NFI = 0.877, CFI = 0.888, and RMSEA = 0.033. Finally, the blended learning model yielded favorable results with CMIN/ DF = 2.59, Chi-square = 138.4, *p*-value = 0.000, GFI = 0.940, NFI = 0.931, CFI = 0.940, and RMSEA = 0.011 (see Table 3).

5.4.2 Convergent and discriminant validity. Convergent validity was measured using the Average Variance Extracted (AVE) and Construct Reliability (C.R.). In face-to-face learning, across two factors, FTF-1 and FTF-2, the AVE values were 0.49 and 0.39, with construct reliability at 0.73 and 0.52, respectively. Online learning exhibited AVEs of 0.52 and 0.67 for OL 1 and OL 2 and constructed reliability of 0.86 and 0.82. The analysis of blended learning revealed that the Average Variance Extracted (AVE) for the constructs BL 1 and BL 2 was

0.48 and 0.46, respectively. Additionally, the construct reliability for BL 1 and BL 2 was found Asian Association to be 0.81 and 0.77, respectively (see Table 4). of Open Further, discriminant validity assessments showed that for all latent variables, the Universities

Further, discriminant validity assessments showed that for all latent variables, the Average Variance Extracted (AVE) exceeded the squared inter-construct correlations, confirming discriminant validity (as shown in Table 4). These results collectively affirm the validity of the constructs used in the questionnaire, emphasizing the robustness of the measurement instruments employed in this study (see Figures 2–4).

## 6. Face-to-face, online, and blended learning impact on students' satisfaction

This study utilized multiple regression analysis to assess the impact of Face-to-Face (F2F), Online, and Blended Learning approaches on learner satisfaction, aiming to inform the development of an effective Blended Learning (BL) model (see Figure 5) for post-COVID Indian classrooms. The model summary reported a correlation coefficient (R) of 0.529, accounting for 53% (approx.) of the variance. The determination coefficient ( $R^2$ ) stood at 0.280, suggesting that these learning modes explain 28% of the variance in learning satisfaction. The adjusted coefficient of determination (adjusted  $R^2 = 0.275$ ) was slightly lower, providing a more accurate representation of the model's fit in predicting satisfaction.

The coefficient table summary highlighted the varying impacts of these learning modalities on learning satisfaction. F2F learning showed a minimal and statistically insignificant effect, contributing only 7% to the variance in learning satisfaction ( $\beta = 0.070$ , *p*-value = 0.119). On the other hand, Online learning had a significant positive effect, explaining 28.3% of the variance in learning satisfaction ( $\beta = 0.283$ , *p*-value = 0.000). Similarly, Blended Learning exhibited a significant positive relationship with learning satisfaction, accounting for 27.2% of the variance ( $\beta = 0.272$ , *p*-value = 0.000). These insights, detailed in Table 5, underscore the differential influence of learning approaches on student satisfaction.

## 7. Discussions and implications

This study aimed to devise an effective blended learning (BL) framework for post-vaccine classrooms in India by investigating face-to-face (F2F) and online learning factors. Guided by two primary questions - "Did we really learn from the COVID-19 virus?" and "Are we really prepared with appropriate pedagogical techniques for post-vaccine education in India?" - we examined university students' perceptions and satisfaction with three learning approaches:

| Face to face learning |                       |                   | Online lea            | rning             | Blended le            | arning            |
|-----------------------|-----------------------|-------------------|-----------------------|-------------------|-----------------------|-------------------|
| Model fit<br>elements | Parsimonious<br>Model | Improved<br>model | Parsimonious<br>model | Improved<br>model | Parsimonious<br>model | Improved<br>model |
| CMIN/DF               | 3.04                  | 2.17              | 3.19                  | 2.81              | 5.87                  | 2.59              |
| $\chi^2$              | 65.9                  | 62.7              | 199.9                 | 196.2             | 177.8                 | 138.4             |
| ĎF                    | 8                     | 7                 | 25                    | 22                | 24                    | 21                |
| <i>p</i> -value       | 0.000                 | 0.000             | 0.000                 | 0.000             | 0.000                 | 0.000             |
| GFI                   | 0.939                 | 0.958             | 0.898                 | 0.919             | 918                   | 0.940             |
| NFI                   | 0.863                 | 0.869             | 0.880                 | 0.877             | 0.911                 | 0.931             |
| CFI                   | 0.875                 | 0.880             | 0.893                 | 0.888             | 0.922                 | 0.940             |
| RMSEA                 | 0.041                 | 0.037             | 0.046                 | 0.033             | 0.053                 | 0.011             |

Table 3. Goodness of fit measures of face-toface, online and blended learning

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learning

online and blended

3

Source(s): Primary data

0.37

FTF2

0.120 1

OL2

| Face-to-face<br>learning | Estimates<br>(Standardized) | <i>p</i> -value | Online<br>learning   | Estimates<br>(Standardised) | <i>p</i> -value | Blended<br>learning  | Estimates<br>(Standardised) | <i>p</i> -value |
|--------------------------|-----------------------------|-----------------|----------------------|-----------------------------|-----------------|----------------------|-----------------------------|-----------------|
| $FTF6 \leftarrow F1$     | 0.831                       | 0.000           | $OL5 \leftarrow F1$  | 0.717                       | 0.000           | BL5 $\leftarrow$ F1  | 0.781                       | 0.000           |
| $FTF7 \leftarrow F1$     | 0.709                       | 0.000           | $OL6 \leftarrow F1$  | 0.752                       | 0.000           | $BL6 \leftarrow F1$  | 0.766                       | 0.000           |
| $FTF8 \leftarrow F1$     | 0.441                       | 0.000           | $OL7 \leftarrow F1$  | 0.699                       | 0.000           | $BL7 \leftarrow F1$  | 0.701                       | 0.000           |
| $FTF1 \leftarrow F1$     | 0.497                       | 0.000           | $OL8 \leftarrow F1$  | 0.652                       | 0.000           | $BL9 \leftarrow F1$  | 0.556                       | 0.000           |
| $FTF3 \leftarrow F1$     | 0.415                       | 0.000           | $OL9 \leftarrow F1$  | 0.781                       | 0.000           | $BL10 \leftarrow F1$ | 0.621                       | 0.000           |
| $FTF5 \leftarrow F1$     | 0.626                       | 0.000           | $OL10 \leftarrow F1$ | 0.744                       | 0.000           | $BL1 \leftarrow F2$  | 0.742                       | 0.000           |
|                          |                             |                 | $OL1 \leftarrow F2$  | 0.875                       | 0.000           | $BL2 \leftarrow F2$  | 0.804                       | 0.000           |
|                          |                             |                 | $OL2 \leftarrow F2$  | 0.821                       | 0.000           | BL3 $\leftarrow$ F2  | 0.666                       | 0.000           |
|                          |                             |                 | $OL3 \leftarrow F2$  | 0.629                       | 0.000           | $BLA \leftarrow F2$  | 0.483                       | 0.000           |

|   | Face-to-face                 |                 | Average<br>variance   | Construct                        |                     | Onlii       | gent validity<br>ne learning<br>Average<br>variance | Construct                             |                        |               | ed learning<br>Average<br>variance |     | onstruct                    |
|---|------------------------------|-----------------|-----------------------|----------------------------------|---------------------|-------------|---|---------------------------------------|------------------------|---------------|------------------------------------|-----|-----------------------------|
|   | Dimensions                   | Items           | extracted             | reliability                      | Dimensions          | Items       | extracted   | reliability                           | Dimensions             | Items         | extracted                          | re  | eliability                  |
|   | Face-to-<br>Face<br>Learning | FTF1<br>FTF2    | 0.49<br>0.37          | 0.73<br>0.52                     | Online<br>Learning  | OL1<br>OL2  | 0.52<br>0.61  | 0.86<br>0.82                          | Blended<br>Learning    | BL1<br>BL2    | 0.48<br>0.46                       |     | 0.81<br>0.77                |
| Table 4.  | Face-to-face                 | learning<br>No. | Average               | Correlation<br>(squared<br>inter | ]                   |             | inant validit<br>ne learning<br>Average             | y<br>Correlation<br>(squared<br>inter |                        | Blende<br>No. | ed learning<br>Average             | (se | relation<br>quared<br>inter |
| Goodness of fit<br>measures, convergent<br>and discriminant | Items of dimensions          | Of<br>items     | variance<br>extracted | construct<br>correlation)        | Items of dimensions | Of<br>items | variance<br>extracted                               | construct<br>correlation)             | Items of<br>dimensions | Of<br>items   | variance<br>extracted              | CO  | nstruct<br>relation)        |
| validity of face to face,                                   | FTF1                         | 3               | 0.49                  | 1 0.014                          | OL1                 | 6           | 0.52  | 1 0.272                               | BL1                    | 5             | 0.48                               | 1   | 0.216                       |

3

0.61

0.522 1

4

0.46

BL2

0.465 1

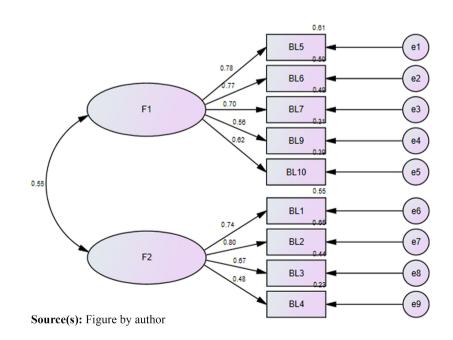
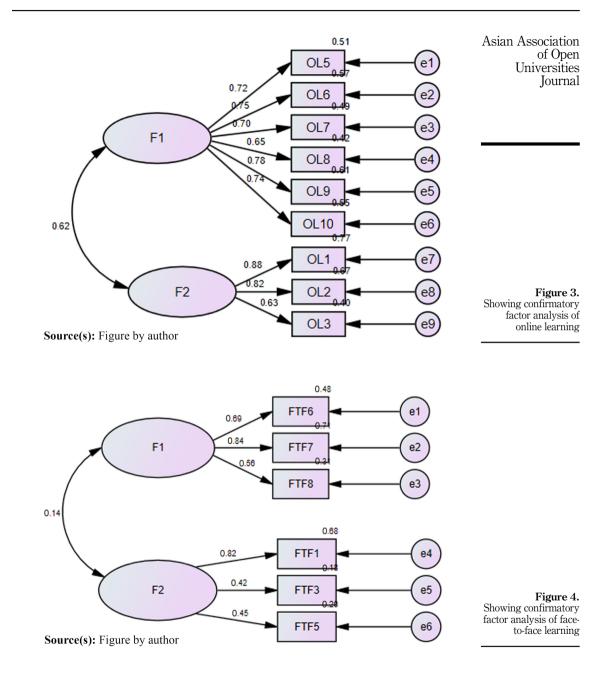


Figure 2. Showing confirmatory factor analysis of blended learning



F2F (pre-pandemic experiences), online learning (during the initial COVID-19 lockdown), and BL (between COVID-19 waves, post-vaccination). The study focused on aspects such as the teacher's role, educational techniques, and technological integration.

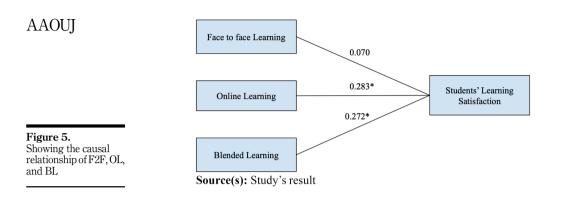


Table 5. Regression

|             | Estimate                            | Error  | Standardised estimates ( $\beta$ )   | t   | Sig. (p-value)   |
|-------------|-------------------------------------|--|--|---|--|
|             | 18.863                              | 2.659  | -  | 7.094   | 0.000  |
| arning      | 0.184                               | 0.118  | 0.070  | 1.564   | 0.119  |
| ζ           | 0.390                               | 0.075  | 0.283  | 5.182   | 0.000  |
| ng          | 0.391                               | 0.073  | 0.272  | 5.339   | 0.000  |
| $R^2$       |                                     | Adjusted $R^2$   | SE of the estimate   |   | Sig. F change  |
| 0.280       |                                     | 0.275  | 7.76392  |   | 0.000  |
| ables: lear | ning satisfac                       |  | ng, online learning, blended lear  | ming  |  |
|             | 0.280<br>bendent va<br>iables: lear | $\frac{18.863}{0.184}$ arning 0.184<br>g 0.390<br>ng 0.391<br><sup>Ty</sup><br>$R^2$<br>0.280<br>bendent variables: face | 18.863         2.659           arning         0.184         0.118           g         0.390         0.075           ng         0.391         0.073           "Y $R^2$ Adjusted $R^2$ 0.280         0.275           bendent variables: face to face learning satisfaction | 18.863         2.659         -           arning         0.184         0.118         0.070           g         0.390         0.075         0.283           ng         0.391         0.073         0.272           TY $R^2$ Adjusted $R^2$ SE of the estimate           0.280         0.275         7.76392           bendent variables: face to face learning, online learning, blended lear         iables: learning satisfaction | 18.863       2.659       -       7.094         arning       0.184       0.118       0.070       1.564         g       0.390       0.075       0.283       5.182         ng       0.391       0.073       0.272       5.339         Y $R^2$ Adjusted $R^2$ SE of the estimate         0.280       0.275       7.76392         bendent variables: face to face learning, online learning, blended learning iables: learning satisfaction |

The findings revealed a positive significant association between learners' satisfaction and both online learning and BL, with  $\beta$  values of 28.3 and 27.2, respectively, and a *p*-value of 0.000. However, F2F learning showed an insignificant effect on satisfaction ( $\beta = 0.070$ , *p*-value = 0.119). These results align with prior studies (Marriot *et al.*, 2004; Alammary, 2022; Ogbona, 2021) that emphasize the importance of F2F learning in fostering a sense of belonging and the teacher's vital role. Additionally, online learning's impact on student satisfaction was found to be greater than that of BL (see Table 5), resonating with the findings of Singh *et al.* (2021), Tratnik *et al.* (2019), Kintu *et al.* (2017), and She *et al.* (2021).

Regression analysis demonstrates online learning modalities significantly impact learner satisfaction, with an *R*-squared value of 0.280 indicating a considerable variance explanation in satisfaction levels. Key benefits of online learning, including superior internet systems, engaging animations, and flexible learning schedules, contribute to enhanced collaboration, knowledge, and skill development. Interactivity within online platforms is pivotal for forging social connections and boosting course satisfaction, as it positively affects academic skills and self-efficacy, aligning with findings from various studies (Demir Kaymak and Horzum, 2013; Strauß and Rummel, 2020; Aragon, 2003; Kim and Frick, 2011; Nelson *et al.*, 2005; Tekin *et al.*, 2014; Gebauer *et al.*, 2020). Teachers play a crucial role in guiding and fostering growth, with interaction with course content crucial for cognitive skill development (Miller and Brickman, 2004; McMahon *et al.*, 2009; Moore, 1989). Regression coefficients for online and blended learning show positive impacts on satisfaction, with standardized estimates of 0.283 and 0.272, respectively, and *p*-values of

0.000, highlighting the importance of technology, pedagogy, interactivity, face-to-face Asian Association interactions, content frequency, self-regulation, software quality, and tools in the blended learning system.

This study enriches existing knowledge by identifying key factors for effectively implementing Blended Learning (BL) in post-vaccine classrooms in India. For this, we have examined the first-order variable against the second-order variables, focusing on the crucial elements of Face-to-Face (F2F), Online Learning (OL), and BL to evaluate their combined effects on students' learning satisfaction. The research findings indicate that among the three learning modalities—face-to-face (F2F), online learning (OL), and blended learning (BL)—it, OL and BL that had a positive influence on learner satisfaction. This is evidenced by significant beta values ( $\beta = 0.283$  for OL and  $\beta = 0.272$  for BL) and p-values (0.000 for both OL and BL). Notably, BL factors were the most influential factors for Indian students, surpassing both OL and F2F in terms of impact. This result holds considerable implications for educational strategies within Indian universities.

This study is among the pioneering efforts to evaluate learners' overall educational satisfaction with Face-to-Face (F2F), Online Learning (OL), and Blended Learning (BL) in India's post-vaccine educational landscape. It provides foundational insights into the factors influencing learning satisfaction within a BL framework. By examining learners' perceptions and experiences with F2F, OL, and BL, the study assesses their collective influence on learning satisfaction. These findings lay the groundwork for future research to uncover additional factors crucial for learning satisfaction in BL settings. The study reveals a preference for OL and BL over F2F, highlighting the importance of blending the teacher's role with online learning elements to enrich the academic experience.

The preference for BL, demonstrated in this study, underscores the need for educators, universities, and other stakeholders to prioritize BL factors. The BL framework combines face-to-face and online teaching, allowing students to clarify doubts from online sessions in face-to-face interactions with teachers. In this model, online sessions are delivered through the university's e-system, while F2F sessions take place on campus, necessitating highquality software and hardware, along with proficient instructors. The study emphasizes the importance of regular software updates, ease of operation by instructors, and training in digital and in-class strategies to ensure smooth course functioning. The study's themes include technological, educational, and interactivity factors.

Dhawan (2020) noted the necessity of computer and technology skills in a BL environment. Our study confirms students' competence in using computers and the internet within BL, diverging from Kintu et al. (2017), who found that reluctance and incompetence in these areas led to higher dropout rates in OL and BL courses. Unlike Cohen (2012) findings, issues like family responsibilities, work timing, accessibility, and connectivity did not significantly hinder participation in our study, as most participants were full-time students in a BL course. Our results align with Qamar et al.'s (2023) research, showing positive student attitudes towards OL and BL, which are viewed as instrumental to educational success. Therefore, university administrations must implement systems that not only impart academic knowledge but also foster creativity, collaborative skills, and selfefficacy, preparing students for efficient future workplace performance.

#### 8. Conclusion, limitations, and future direction

This study concludes that successful blended learning (BL) implementation in post-vaccine classrooms necessitates the careful integration of technology in both teaching and learning. A thorough examination of face-to-face (F2F) and online learning elements is essential to develop an effective BL framework. Key indicators from online and blended learning

significantly influenced learners' satisfaction. Out of twenty-nine variables in the measurement model (CFA), twenty-four were relevant across the latent variables of F2F, online, and blended learning. Nine variables, each from online and blended learning, significantly contributed to understand the causal relationship with learners' satisfaction. However, six F2F variables showed an insignificant causal link, indicating a need for further research to explore their impact on different educational settings.

Every research permeates through limitations, and this study is no exception. This work gathered its data solely from students at two prominent universities in India-one public and the other private. This sampling does not necessarily capture the diversity of the broader student population. Therefore, generalizing these findings to all Indian university students should be done with caution, considering the limited sample size, distinctive characteristics of the learners, and the specific educational infrastructures of the surveyed institutions. Future research would gain from a broader and more varied sample spanning multiple universities to attain more widely applicable findings. Moreover, this study's use of purposive sampling was specifically chosen to investigate causal relationships between study variables, which may not be as broadly generalizable as those obtained through longitudinal and experimental research designs. Future inquiries are warranted considering factors such as students' perceptions, the attributes of teachers, and the emotional components of blended learning (BL). While this study has focused on quantitative analysis, subsequent research using qualitative methodologies could also provide a deeper, more nuanced comprehension of BL, particularly in addressing aspects not explored in the present study.

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