The Validation of the Information Systems Success Model: LMS Integration during Covid-19

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Abstract

This study adapted DeLone & McLean information system success model (D&M IS success model) regarding the implementation of learning management system (LMS) during Covid-19. Six variables are included; system quality, information quality, service quality, system usage, user satisfaction, and net benefits with 23 initial items. A total of 279 undergraduate students from a public university in Indonesia participated in this study. The factor structure of the instrument was investigated using a survey design. The survey data were calculated using exploratory factor analysis (EFA) and confirmatory factor analysis (CFA). Six variables emerged from the EFA methods, establishing a valid and reliable model; a few items were eliminated due to cross-loading. The suggested model was successfully mapped as a consequence of the results. The CFA confirmed that the instrument was suitable for the Indonesian setting. The findings led to the development of 19 reliable and valid items. The measured scale has psychometric qualities, allowing for future research with a tool to capture D & M IS success model technology integration.

Keywords
CFA, EFA, IS success model, Validation

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Introduction

The use of technology has transformed traditional teaching and learning techniques into a more active and dynamic condition. Face-to-face contact is no longer the sole way for instructors and students to communicate (Hernandez-de-Menendez et al., 2020). In current years, the use of technology in education, particularly in higher education, has increased student access and engagement. Since then, the educational methodology has evolved from traditional techniques and toward electronic learning (e-learning). The learning management system (LMS) is one type of e-learning used in higher education. The demand for quick transmission of knowledge and information at any time and from any location has fueled the rise of LMS throughout the world and is a critical component of corporate success of educational activities. When combined with the prevalence of coronavirus disease 2019 (Covid-19), LMS is a recipe for the disaster (Abazi-Bexheti et al., 2018; Rossini et al., 2021). LMS is a web-based learning platform that allows users to access information and knowledge regardless of time or location. LMS is a comprehensive e-learning platform with full multimedia integration, instructor-led and real-time instruction, and a collaborative environment. The real-time and synchronous LMS distributions and the asynchronous LMS distributions are the two forms of LMS (Namada, 2021). It has become an essential component of remote learning; an idea that must be fully grasped for the new learning environment paradigm.

Humanity has encountered numerous challenges since the Spanish flu outbreak. Without question, the Covid-19 is unparalleled. With almost three billion individuals under quarantine since its start, the scope and impact are unprecedented in contemporary global history. The technology, on the other hand, substantially separates the contemporary situation from the past, altering quarantine history. The employment of technology in all parts of life is unavoidable in this Covid-19 environment (Gabr et al., 2021). Several studies have looked at how technology is used in schools during the pandemics. However, there are few empirical studies on the success of LMS implementation during Covid-19. The current study is part of a larger attempt to develop a valid and reliable scale for measuring LMS implementation using the D&M IS success model. The research was carried out at one university, where students utilized LMS on a regular basis during the Covid-19 teaching and learning process.

Literature Review

Understanding information system success is a topic that many scholars, practitioners, and management stakeholders are interested in. This knowledge helps to emphasize the system's worth and may be used to inform future decisions about similar systems. There are several methods to measure success; the D&M IS model is among the widely used and well-validated metrics. The model was initially proposed in 1992; it was revised in 2003 (DeLone & McLean, 2003). The model facilitates six interrelated dimensions: system quality, information quality, service quality, system usage, user satisfaction, and net benefits (Delone & McLean, 2014; DeLone & McLean, 2003; Wang, 2008). In previous
research, the model has been used and verified; e-commerce platforms (Sharma & Aggarwal, 2019; Tam et al., 2020), knowledge sharing (Halonen et al., 2010; Sarkheyli & Song, 2019), e-government (Lessa & Tsegaye, 2019; Mellouli et al., 2020), and technology integration in education (Al-Azawei, 2019; Safsouf et al., 2020; Shahzad et al., 2021). It’s worth noting that the majority of research using the D&M IS model to measure success have been conducted in developed nations, with just a handful specifically confirming the model regarding LMS integration in developing ones, particularly during Covid-19 (Wang, 2008). As a result, the primary goal of the current research is to evaluate the proposed D&M IS model to evaluate LMS integration success during Covid-19 in Indonesia.

The D&M IS model, which presents six interconnected constructs of information system success metrics, is used to drive this research (Al-Azawei, 2019; Halonen et al., 2010; Lessa & Tsegaye, 2019; Mellouli et al., 2020; Safsouf et al., 2020; Sarkheyli & Song, 2019; Shahzad et al., 2021). 1) System quality is said to assess an information system’s desired qualities. This has been assessed in several information systems research utilizing factors such as system tools, reaction speed, and adaptability. This research, on the other hand, evaluated system quality by looking at the ease of use of LMS integration and their function and adaptability. 2) Information quality is content concerns and output qualities of information systems. It has been assessed by evaluating an information system’s output about timeliness, correctness, dependability, and trustworthiness. 3) Service quality is determined by the level of assistance provided by the creator of the information system. Service quality characteristics like assurance and response by the systems support department, and the provision of user training, have been used in studies to measure this. Service quality was assessed in this study by looking at the technical assistance provided to users of a hospital information system, the network infrastructure in place, and the system’s dependability. 4) Intention to use is focused on evaluating how an information system is utilized. Various studies have looked at actual usage or, in certain cases, frequency of use to determine this. 5) User satisfaction is one of the most significant indicators of a system’s performance, and it’s frequently assessed by total user happiness. 6) Net benefits: this is also one of the most significant indicators of information system performance, as it indicates how much an information system contributes to the success of various stakeholders, whether good or bad. It has been assessed by measuring individual or organizational influence on occasion.

**Need for an instrument**

Statistical data on technological integration, as seen by teachers, has been the subject of many studies (Dong et al., 2015; M. Liu, 2013; Liu, 2013; Ndongfack, 2015; Polly et al., 2010). Besides, students have also been involved in the context of technology integration research (Al-Ani, 1979; Dasig & Pascua, 2016; Ervin, 2014; Lisenbee & Ford, 2018). However, just a few research provided enough data on success of information technology (DeLone & McLean, 2003; Halonen et al., 2010; Sarkheyli & Song, 2019; Wang, 2008), especially in developing countries. Therefore, this study’s objective is to examine the validity and reliability of the instrument in the context of developing country perceived by students regarding the success of the implementation of LMS during Covid-19. Instrument
development should include a sufficient number of indicators to fit the setting and context (Connell et al., 2018; Hosseini & Kamal, 2012; Jamieson-Proctor et al., 2013; Valtonen et al., 2015; Zelkowski et al., 2013). It aims at capturing critical aspects of constructs of the study. This study refers to the context and setting in Indonesian scope through the use of Exploratory Factor Analysis (EFA) and Confirmatory Factor Analysis (CFA) (James, 2009; Kyriazos, 2018; Padgett & Morgan, 2021). Specifically, the study was established by the following research question: How valid and reliable the proposed instruments regarding LMS implementation success among Indonesian students during Covid-19?

**Methodology**

**Research design and participants**

This study applied survey as the main data collection method (Ball, 2019; Geldsetzer, 2020; Weiss et al., 2016). We initiated the survey instrument by thoroughly evaluating related previous studies (Andrews & Diego-Mantecón, 2015; Hanniball et al., 2021). Afterwards, the instrument indicators was validated through content validity and distributed for a pilot study (Hazzi & Maaldaon, 2015; Leon et al., 2011). After the validation of the data normality (Alejo et al., 2015; Miot, 2017; Noél, 2021), the data were assessed for the validity and reliability through exploratory factor analysis (EFA) and confirmatory factor analysis (CFA). The present study's population was made up of all undergraduate students at an Indonesian public university. Cluster random sampling was used by the researchers because it allows them to investigate the selection of groups rather than individuals. Three hundred undergraduate students were given online survey instruments, and a total of 279 students responded. To gain a better comprehension of the data, the instruments were prepared in the participants’ native language, Indonesian.

**Data collection**

A questionnaire was developed to gather the data required for analysis. In ensuring validity, 24 items were adapted from prior studies with validated scales (Cho et al., 2015; Ojo & Popoola, 2015; Pai & Huang, 2011; Tilahun & Fritz, 2015). We discussed the questionnaire with five experts and five users in order to assess all items through the face and content validity process. Adjustments were made as needed and 21 items were submitted for the primary data collection; three were removed because they were inappropriate for the study’s topic, context, and setting. The current study employs a survey design consisting of methods for quantitative approaches that allow surveyors distribute an instrument to sample or the entire population to gather information on the respondents’ views and perceptions. Individual accounts of social reality make it easier to build a foundation for cultivation and silent study. Survey research is very critical in education. A survey is described as a combination of data obtained from the responses of a sample of people (Liang et al., 2013; Williams, 2013). Creswell (2014) mentioned that the survey design contrasts from the experimental design in that it does not provide a treatment for the participants or subjects. Because investigators do not change the settings like experimental researchers do, they are
unable to draw conclusions about cause and effect. Instead of informing robust interpretations, surveys exaggerate data patterns (Rowan et al., 2001). In a quantitative, survey research is a technique in which researchers disseminate a survey data collection device to a sample or the entire population in order to elucidate their views, beliefs, behaviors, or traits (Chan, 2012).

Data analysis

The current study covered several data screening concerns, namely missing data, multicollinearity, outliers, and normality, before moving on to the primary data analysis. The statistical results were computed using SPSS 23.0. For each variable, we used a box plot to calculate outliers. The skewness and kurtosis were assessed to ensure that the data was normal (-1.96 to +1.96 at the 0.05 significance level) (Chou et al., 1998; Singh & Masuku, 2014). When the correlation matrix was more than the cutoff of .900, multicollinearity was detected. Following that, data was examined in two phases for each construct: EFA and CFA. The kaiser-meyer-olkin (KMO) value, bartlett’s value, factor loading, eigenvalue, scree plot, and varimax rotation were all included in the EFA. The KMO index needs to be higher than .500. A KMO score of less than .500 indicates that the sample size for the EFA technique is insufficient; the results may not be trustworthy. At p .050, bartlett’s test of sphericity was significant. Each indicator’s factor loading should be more than .500 (Habibi et al., 2020; James, 2009).

An eigenvalue is the proportion of variance contribution retrieved by each factor using factor analysis; an eigen of less than 1.00 must be eliminated, and communality must be > .300. According to CFA, quality of fit was assessed using some measurements, such as loading of >.500 (Kyriazos, 2018; Truong et al., 2010; Zainudin, 2012), chi-square at p > .05, Comparative Fit Index or CFI of > .800, and Root Mean-Square Error of Approximation or (RMSEA of < .080) (Padgett & Morgan, 2021); details are presented in Table 4. The stability of the values acquired is characterized as reliability. Cronbach’s alpha, Construct Dependability (CR), and Average Variance Extracted (AVE) values were computed to determine the data’s reliability. Cronbach’s alpha must be larger than .700, CR values should be greater than .600, and AVE values should be greater than .500.

Findings

Preliminary analysis

The quantity of missing data in the current study ranged from 0% to 0.4 percent on each item. Multiple imputations, an iterative type of stochastic imputation, were employed to cope with the missing data. The observed data distribution was used to estimate several values that reflected the real value’s uncertainty instead of replacing it with a single value. Table 1 shows the correlation matrix, skewness, and kurtosis, indicating a satisfactory study. The early study of system quality, information quality, service quality, system utilization, user happiness, and net benefits found that they were all univariate normal (skewness and kurtosis values ranging from -.419 to -.132 and from .137 to 1.193, respectively).
Inter-correlations among the six variables ranged from .295 to .615 in terms of multicollinearity; discriminant validities were achieved (correlations of .900).

**Table 1. Correlation matrix, skewness, and kurtosis**

|                  | SQ    | IQ    | SerQ  | SU    | USat  | NB    |
|------------------|-------|-------|-------|-------|-------|-------|
| System quality   | 1     | .615**| .475**| .465**| .393**| .436**|
| Information quality | .615**| 1     | .482**| .495**| .354**| .455**|
| Service quality  | .475**| .482**| 1     | .418**| .360**| .392**|
| System usage     | .465**| .495**| .418**| 1     | .457**| .379**|
| User satisfaction| .393**| .354**| .360**| .457**| 1     | .295**|
| Net benefits     | .436**| .455**| .392**| .379**| .295**| 1     |
| Mean             | 3.6254| 3.5968| 3.4719| 3.7145| 3.6476| 3.7195|
| SD               | .45867| .53941| .62578| .58705| .61913| .62686|
| Skew             | -.318 | -.154 | -.318 | -.419 | -.214 | -.132 |
| Kurt             | .137  | 1.145 | 1.121 | 1.193 | 1.159 | 1.137 |

**Exploratory factor analysis**

The EFA included the measurement of all 21 items. One item (SQ1, “The system is easy to use”) with a loading value of below .500 was dropped. KMO values were above .500, and all values were significant at $p < .005$ (Table 2). With an eigenvalue of greater than 1.0, six factors reported: system quality (1.024), information quality (1.590), service quality (1.377), system usage (1.377), user satisfaction (1.936), and net benefits (6.937). The factor loadings of all variables are above .400 system quality (.482 to .769), information quality (.539 to .745), service quality (.727 to .809), system usage (.585 to .773), user satisfaction (.759 to .813), and net benefits (.681 to .747). All communality values exceed .300, showing sufficient values of communality. Table 3 informs all values of loadings, communalities, eigenvalues, cross-loadings.

**Table 2. KMO and Bartlett’s test**

| Kaiser-Meyer-Olkin Measure of Sampling Adequacy. | .865 |
| Bartlett’s Test of Sphericity                   |     |
| Approx. Chi-Square                             | 2585.691 |
| df                                             | 190 |
| Sig.                                           | .000 |
Table 3. Loadings, communalities, eigenvalues, cross-loadings

| Variable                  | Eigen. | Item (statement) | M      | SD     | Comm. 1 | 2    | 3    | 4    | 5    | 6    |
|---------------------------|--------|------------------|--------|--------|---------|------|------|------|------|------|
| Net benefits              | 6.937  | NB3              | 3.6237 | .72341 | .747    | .809 |
| NB1                       |        |                  | 3.8136 | .84459 | .681    | .764 |
| NB2                       |        |                  | 3.7885 | .74071 | .701    | .740 |
| NB4                       |        |                  | 3.6523 | .79403 | .669    | .738 |
| User satisfaction         | 1.936  | Usat2            | 3.6344 | .68570 | .759    | .859 |
| Usat3                     |        |                  | 3.6631 | .68470 | .764    | .833 |
| Usat1                     |        |                  | 3.6452 | .72437 | .813    | .818 |
| Information quality       | 1.590  | IQ2              | 3.6631 | .67411 | .745    | .818 |
| IQ1                       |        |                  | 3.5305 | .69295 | .714    | .744 |
| IQ4                       |        |                  | 3.5197 | .69334 | .539    | .644 |
| IQ3                       |        |                  | 3.6738 | .67141 | .605    | .553 |
| Service quality           | 1.377  | SerQ3            | 3.5878 | .69840 | .809    | .834 |
| SerQ2                     |        |                  | 3.3154 | .70518 | .747    | .803 |
| SerQ1                     |        |                  | 3.5125 | .76268 | .727    | .750 |
| System usage              | 1.240  | SU3              | 3.7706 | .71298 | .773    | .812 |
| SU2                       |        |                  | 3.5161 | .72389 | .706    | .788 |
| SU1                       |        |                  | 3.8566 | .70524 | .585    | .616 |
| System quality            | 1.024  | SQ4              | 3.5699 | .53774 | .769    | .844 |
| SQ3                       |        |                  | 3.5806 | .52947 | .767    | .836 |
| SQ2                       |        |                  | 3.7240 | .67789 | .482    | .499 |

Confirmatory factor analysis

CFA was conducted for the factorial validity of the six variables. The measurement model was satisfactory after one covariance activity was done between IQ1 and IQ3. One item was dropped since the value of loading is less than .500 (NB4, the system is an important and valuable aid to me in the performance of my classwork) (Truong et al., 2010; Zainudin, 2012). All loading values for the CFA surpass the standard cutoff value after the dropping. 500 (Figure 1). Figure 1 also shows the standardized coefficients of the CFA, addressing the correlation between factors and items for all variables: $\chi^2/df = 1.830$, CFI = .800, and RMSEA = .055. This computation was exceeding the threshold value of .500 (Table 4). The following six popular model-fit metrics were used to assess the model's goodness-of-fit: The chi-square ratio, goodness-of-fit index (GFI), adjusted goodness-of-fit index (AGFI), normalized fit index (NFI), root mean square residual (RMSR), and root mean square error of approximation (RMSEA) are all terms that can be used to describe how well a model fits. Table 4 indicates that the model fit indices exceeded their suggested acceptability thresholds, indicating that the measurement model fits the data quite well.
Table 4. *model fit indices*

| Parameter                      | Threshold | Obtained value |
|--------------------------------|-----------|----------------|
| Chi-square ratio ($\chi^2/df$) | ≤3.406    | 1.830          |
| GFI                            | ≥0.90     | .906           |
| AGFI                           | ≥0.80     | .868           |
| CFI                            | >0.80     | .800           |
| RMSR                           | ≤0.10     | .065           |
| RMSEA                          | ≤0.08     | .055           |

Figure 1. *CFA results*

Cronbach’s alpha and composite reliability (CR) values of all variables were found to be satisfactory of above .700: system quality (CR = 0.772; $\alpha = .706$), information quality (CR = 0.814; $\alpha = .799$), service quality (CR = 0.830; $\alpha = .833$), system usage (CR = 0.767; $\alpha = .760$), user satisfaction (CR = 0.853; $\alpha = .864$), and net benefits (CR = 0.814; $\alpha = .786$). In addition, the AVE for all variables also exceed the desirable threshold value of .500, denoting that this study had acceptable discriminant validity (Table 6). Through the examination and elaboration of EFA and CFA in measuring the scale, the finding elaboration suggested the establishment of the validity and reliability of the survey instrument, a scale to provide a valid and reliable scale to measure the success of LMS implementation through the use of D&M IS success model.
### Table 5. CFA results of all constructs

| Variable          | Items                                                                 | Loading | CR   | AVE   | α     |
|-------------------|-----------------------------------------------------------------------|---------|------|-------|-------|
| Net benefits      | NB3 (Overall, the system is successful)                               | .720    | 0.814| 0.770 | .786  |
|                   | NB1 (The system has a positive impact on my learning)                 | .820    |      |       |       |
|                   | NB2 Overall, the performance of the system is good                    | .770    |      |       |       |
| User satisfaction | Usat2 (I think the system is very helpful)                             | .760    | 0.853| 0.891 | .864  |
|                   | Usat3 (Overall, I am satisfied with the system)                       | .900    |      |       |       |
|                   | Usat1 (I have a positive attitude or evaluation about the way the system functions) | .900    |      |       |       |
| Information quality| IQ2 (Information I get from the system is accurate)                   | .740    | 0.814| 0.723 | .799  |
|                   | IQ1 (Information from the system is relevant to my work)              | .780    |      |       |       |
|                   | IQ4 (The information is presented in a useful format)                 | .630    |      |       |       |
|                   | IQ3 (It is easy to understand information from the system)            | .740    |      |       |       |
| Service quality   | SerQ3 (Overall, the support services meet my needs)                  | .810    | 0.830| 0.787 | .833  |
|                   | SerQ2 (The support services give me individual attention)             | .740    |      |       |       |
|                   | SerQ1 (The support services for the system are dependable)            | .810    |      |       |       |
| System usage      | SU3 (I only use the system when it is absolutely necessary for learning) | .770    | 0.767| 0.723333| .760  |
|                   | SU2 (I depend upon the system)                                        | .710    |      |       |       |
|                   | SU1 (I frequently use the system)                                     | .690    |      |       |       |
| System quality    | SQ4 (I can retrieve information I need easily)                        | .750    | 0.772| 0.723333| .706  |
|                   | SQ3 (The system is easy to learn)                                     | .850    |      |       |       |
|                   | SQ2 (The system is useful)                                            | .570    |      |       |       |

**Discussion**

The process of developing the scale of the current study was conducted within some stages. It aims at producing a scale with examined validity and reliability. Initially, twenty-four items were established, adapted from prior studies (Cho et al., 2015; Ojo & Popoola, 2015; Pai & Huang, 2011; Tilahun & Fritz, 2015). To filter the instruments, three items were removed from the list during face and content validity processes. The items (n. 21) were distributed to respondents (279 undergraduate students) from one public university in Indonesia, EFA was conducted using Varimax rotation with principal component analysis. Through this process, one item was dropped. The dropping indicator process did not result
in removing any important content of the scale. The dropping process helped improve the reliability and validity of the scale, leaving the scale with twenty items for the CFA process. The processes for determining the validity and reliability of an instrument development should contain a sufficient number of items that are appropriate for the setting and context. As a result, it may be able to capture important features of the structures in a study (MacLeod et al., 2018; Maul, 2017).

The technique aims to reveal the unknown elements that impact the co-variation of various LMS integration observations in the Indonesian setting. The technique is then repeated using CFA to confirm the factorial validity of the D&M IS success model D&M IS success model D&M IS success model D&M IS success model D&M IS success model D&M IS success model components in relation to the usage of LMS during COVID-19. The EFA-based data were calculated for CFA using SPSS AMOS 23.0; just one item was eliminated since its loading value was less than its cutoff value. The reliability of the remaining 20 indications was investigated. Cronbach’s alpha, AVE, and CR values are adequate in this procedure, resulting in a legitimate and trustworthy scale. Previous research using similar techniques has looked at the CFA process to corroborate the EFA results. This method is critical for determining sub-construct measurements that are compatible with our knowledge of their nature (Kyriazos, 2018; Miller & Pellegrino, 2018). The scale is appropriate for the measuring model and can help future researchers perform comparative studies.

Conclusion

During the Covid-2019 pandemic, the current study seeks to create and validate the D&M IS success model in the setting of a learning management system (LMS) in a university in a poor nation (Covid-19). The final scale consists of 19 elements divided into six constructions (system quality, information quality, service quality, system usage, user satisfaction, and net benefits). The measured scale has adequate psychometric characteristics and can be used in future research. The scale’s reliability and validity were only tested at one university. It’s also necessary to incorporate a broader range of samples in research and to evaluate the link. Other study settings and contexts were also suggested. Furthermore, during COVID-19, more specific and applicable definitions of extended constructs and sub-constructs for measuring LMS integration will aid in the development of more consistent and exact survey instruments. Valid and trustworthy indications linked to technology integration might be included in such instruments during a future outbreak emphasized by a similar idea. Furthermore, more relevant definitions of technological integration during a pandemic epidemic are required, both conceptually and practically. Mapping the technology use in this context will move the focus away from the method of instruction and onto the subject.

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