

Factors Influencing Student Satisfaction towards Distance Learning Apps During the Coronavirus (Covid-19) Pandemic in Malaysia

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To Link this Article: http://dx.doi.org/10.6007/IJARPED/v10-i2/9546

DOI:10.6007/IJARPED/v10-i2/9546

Published Online: 28 April 2021

Abstract

The latest coronavirus epidemic has witnessed a catastrophic fall in traditional education system and a significant change to app-based learning to maintain learning continuity. The conventional face-to-face teaching approach has been drastically altered with the addition of a new dimension in which student satisfaction is paramount. This current research paradigm is based on the TAM model. Students residing in Klang Valley, Malaysia, obtained a total of 483 data. By deploying SPSS and AMOS version 24.0, the study found self-efficacy as one of the deciding factors of student satisfaction; there is a strong correlation between self-efficacy and satisfaction. Furthermore, perceived usefulness and perceived ease of use are also influenced by self-efficacy. In contrast, perceived ease of use does not affect perceived usefulness, and perceived usefulness does not affect satisfaction. Finally, to increase the level of distance learning during COVID-19 outbreak, recommendations are made to enhance the virtual learning environment.

Keywords: AMOS, Covid-19, Distance Learning, Malaysia, Student Satisfaction, TAM.

Introduction

Student satisfaction has widely being studied but in distance learning environment it has generated concern due to its impact on the usefulness of teaching and instructional materials. Level of expectation and performance are the primary measurement of satisfaction but cultural differences affect the degree of satisfaction of students about their perception of the services provided (Tian & Wang, 2010). Student satisfaction is an indicator of the quality of an educational program and is regarded as an important factor in the completion of the course (McQuillan, 2010). In most cases, students in postsecondary education programs end up leaving their studies because of unhappiness with their courses (Khan & Iqbal, 2016). Satisfaction among students creates self-esteem that helps students build confidence, find additional skills and gain knowledge in a virtuous cycle. As we see in the world today, the

COVID-19 pandemic is pushing academic institutions like universities to move very quickly to distance learning. We are already in a state of emergency and need to react with various and fully accessible learning methods, such as e-learning systems and virtual learning applications. Studying student satisfaction in the form of distance learning has received attention due to its impact on the efficacy of teaching and educational materials. Distance learning on the other hand has gained popularity during the outbreak of Coronavirus (COVID-19) by shifting away from classroom to online learning in several areas of the globe (Li & Lalani, 2020). Private universities and colleges in Malaysia no longer communicate students face to face and therefore have decided to switch to distance learning to arrest Covid-19 from spreading (Arumugam, 2020). With the implementation of the Movement Control Order (MCO), conventional exams are now replaced by other types of evaluation to maintain the social distance (Haque, Karim, Kabir, & Tarofder, 2020). Distance learning tools in fact are playing a vital role during this pandemic. Distance learning systems can help learner control, schedule, execute and monitor the learning and teaching process. Microsoft Teams, Moodle, WhatsApp, Blackboard, Open Learning Network, WizIQ, Zoom, Facebook Livestream, Google Classrooms, Google Suites and Google Meet were amongst the industry's leading tools and innovations used by educators. Classes were usually performed according to existing timetables to ensure that a sense of "normality" survives, whereas alternative arrangements for evaluations were established.

Since the effectiveness of the distance learning system relies on the willingness and acceptance of the students to use this system, a lack of distance learning system use hinders the realization of benefits (Almaiah & Alismaiel, 2019). Work on this subject is still in its infancy, where the students' viewpoints are not thoroughly investigated. Moreover, studying satisfaction among students regarding to use technology are in need. Studying adoption of distance learning will allow universities to understand the needs of their students better and ultimately lead to an effective distance learning program (El-Masri & Tarhini, 2017). Although distance learning system has been implemented in various universities but thorough analysis on students satisfaction of using particular system is yet to be done during COVID-19 pandemic. Moreover, to perform the specific tasks, ability of using technology is indeed important for students. Additionally, confidence among students in terms of using technology can lead satisfaction. Although technology has widely being accepted in digital learning, students from different study fields are still unaware of using learning technology, thus dissatisfaction correlates. Online education can be efficacious in technologically advanced countries, but it is ineffective in developing countries like Malaysia due to self-efficacy (Adedoyin & Soykan, 2020).

The primary objective of this study is to examine the satisfaction among students regarding to use distance learning application during the coronavirus pandemic in Malaysia. The specific objective of this study is to apply three variables technology self-efficacy, perceived usefulness and perceived ease of use to determine students' satisfaction towards distance learning apps. The remaining paper exhibits five sections including review of literature, research methods, results and findings, and limitations as well as future directions of research.

Literature Review

Successful information systems are often being specified as a precursor of users' satisfaction (Almaiah, 2018). Thus, in distance learning system context, one of the main criteria for continuous success is students' acceptance. Several previous studies have identified an issue

related to distance learning adoption in many countries of the world. Technology plays a vital role which assesses the level of satisfaction of students. By applying TAM Almaiah and Man (2016) critically evaluated the factors which affect the distance learning system. The result articulated the significant role of usability, responsibility, design and functionality which plays an important role to describe the conviction among Malaysian students. Almasri (2018) embeded TAM model to investigate the process of mobile learning among jordanian students. The result showed the significant association between perceived ease of use and individuals attitude (IA), perceived usefulness and IA and interaction and IA. Al-Azawei et al. (2017) investigated the diferencences and perceptions of an individual by deploying TAM model. The study also explored the relationship between technology adoption and satisfaction. Based on TAM3 Al-Gahtani (2016) found that, significant deteminants of technology-assisted learning system enhanced students towards acceptance. Baker et al. (2010) applied TAM 2 for crosscultural evaluation of IT adoption. Findings from the study elicits the adoption of IT among Saudi Arabian organizations. Even though there are multiple studies on e-learning adoption, the present research aims to contribute in the existing literature on investigating the major obstacles and factors of students distance learning satisfaction in a modern context, which is Malaysia, which might also set an example for other underdeveloping nations.

Technology Self-Efficacy (TSE)

Self-efficacy is described as one's judgment about how well different courses of action can be carried out in various prospective circumstances with many unforeseeable and challenging elements (Bandura & Schunk, 1981). The prior research concentrated more on the technical approach of self-efficacy in online learning, e.g. computer self-efficacy (Jan, 2015; Pellas, 2014; Wu, Tennyson, and Hsia, 2010), internet self-efficacy (Lin et al., 2013; Kuo et al., 2014), self-efficacy in the information seeking behavior (Tang & Tseng, 2013) and learning management self-efficacy (Martin, Tutty, & Su, 2010). In the process of learning, self-efficacy conveys evaluations of students on their cognitive capacity rather than affective to meet a particular learning mission or specific goals (Schunk, 1991). However, in the context of internet usage, it signifies the user's ability to perform or carry out the task required. According to Bhattacherjee (2002), the desire of consumers to use a technology is motivated by the satisfaction they have obtained from past experience while using the electronic system. Customer satisfaction was previously studied in the electronic banking context where Rusu and Shen (2012) described computer self-efficacy and convenience as two major factors driving customers to use the electronic banking platform. The introduction of self-efficacy as an essential form of motivation provides a deeper and clearer explanation of why as well as how the technology is being used (Ahmad et al., 2010) but to validate the extended technology acceptance model (ETAM) the study of Islam (2020) failed to produce a significant path-coefficient between computer self-efficacy and sastisfaction. The study futher needs to be rexamined during the outbreak of COVID-19.

In terms of its effect on perceived ease of use (PEOU), computer self-efficacy (CSE) plays a critical role (Madorin & Iwasiw, 1999). A meta-analysis conducted by Abdullah and Ward (2016) posited that, self-efficacy positively and significantly affects PEOU. Similarly, Chen (2012) found that usability of e-portfolios as a part of information technology (IT) are truly based on the ability of students. Different findings are documented regarding the relationship between SE and perceived usefulness (PU) literature. While several studies (Purnomo and Lee, 2013; Mohamed and Karim, 2012; Abbad, Morris and De Nahlik, 2009) have identified a lack of significant correlation between the SE and PU, some studies have found the positive

association between SE and PU (Al-Ammary et al., 2014; Park, 2009; Al-Mushasha, 2013). Chen (2012) also found the positive and significant association between SE and perceived usefulness (PU). Based on the discussion above the proposed hypotheses are as follows:

H1: There is a positive association between TSE and Satisfaction.

H2: There is a positive association between TSE and PU.

H3: There is a positive association between TSE and PEOU.

Perceived Ease of Use (PEOU)

Perceived ease of use (PEOU) is the degree to which a person is of the opinion that using a particular system would be effortless (Davis, 1989). It is hypothesized by Theory of Reasoned Action (TRA) that, behavioral actions are influenced by cognitive factors that can be assessed by predicting behavioral intentions. In addition, the Technology Acceptance Model (TAM) theorizes that the desire of individuals to use technology is measured by their perceived ease of use and perceived usefulness. Almarashdeh and Alsmadi (2016) postulated distance learning as a structured education program in Saudi Arabia where its usable modules are directly influencing the usefulness of the system. Al-Azawei et al. (2017) examined the elearning satisfaction and intention among Iraqi students, whereby, it describes the positive association between PEOU and PU. Sun et al (2008) have investigated design dimention as one of the six dimensions of an integreted model from which perceived ease of use found to have significant influence on e-learners satisfaction. However, Al-Azawei et al (2017) also investigated an intregration model consists of satisfaction, where, PEOU found to have significant influence on satisfaction. But the study of Calli et al (2013) found that satisfaction among students regarding to use e-learning did not influenced by perceived ease of use. Alhawari and Mouakket (2010) also failed to prove the contribution of PEOU on e-satisfaction. Nagy (2018) applied TAM to portray learning satisfaction through video demonstration. The result shows the usability of supplementary application enhance student towards use of the system, because students perceive that system will be useful only if usability is associated. Due to the COVID-19 outbreak, the situation regarding to adopt distance learning might have taken place as several countries have banned public gatherings (Owusu-Fordjour et al., 2020). It is suggesting to academicians to promote and imply online programs because of its easiness and usefulness (Wingo, Ivankova, & Moss, 2017). With that abovementioned literature, the following hypotheses are formed:

H4: There is a positive association between PEOU and PU.

H5: There is a positive association between PEOU and satisfaction.

Perceived Usefulness (PU)

Perceived usefulness is a degree to which a person perceives that the use of an information system will strengthen the efficiency of the task (Davis , 1989). In the context of e-learning system TAM has been widely adopted and examined the relationship between perceived usefulness and users' satisfaction. In this regard, Lee (2010) examined web-based learning system where, perceived usefulness of e-learning found one of the strongest determinant of students satisfaction. Alsabawy et al. (2013) postulate an integral method for controlling and implementing distance learning system, one of those are usefulness. The study also surveyed students persuing degrees in Australian university, result elicits perceived usefulness as strongest predictor of satisfaction and found significant relationship in between. Ibrahim et al. (2019) reveal the key elements of students satisfaction regarding to use technology for learning. Result highlighted that e-learning technology is beneficial and useful for higher

educational institutions and students were satisfied using technology based learning system because of its effectiveness. Useful information system is crucial because it motivates users to engage with particular technology. Hence learning through technology has widely being adopted among university students and satisfaction perceived by them are truly based upon the quality or fact of being useful (Shah & Attiq, 2016). Based on the discussion above, the following hypothesis is proposed:

H6: There is a positive association between PU and satisfaction.

Proposed Research Framework



Figure 1: Conceptual Framework

Research Methodology Construct Measurement

Current research include four constructs namely, technology self-efficacy, perceived usefulness, perceived ease of use and student satisfaction. Measurement items of all the constructs were mainly adopted from previous studies and then adapted to ensure that each element is suitable for the study. In this research, survey questionnaires were adopted primarily from Islam (2020) and Lee (2006). The first four items of technology self-efficacy and three items of student satisfaction were adapted from Islam (2020). Four items of perceived usefulness and perceived ease of use were adopted from Lee (2006). A total of 15 items from those four constructs were examined. To express the degree of acceptability, all measuring items were assessed using five-likert scale (1= strongly disagree to 5= strongly agree). Widely spoken language English was used to design the questionnaire. Last part of the questionnaire was designed to know the respondents age, gender, education and ethnicity.

Data Collection

The target population of this study was university students pursuing their degrees in Malaysia. Right after the movement control order (MCO) lifted researcher has selected Klang Valley for sample collection. Since the number of university students were infinite, total of 483 questionnaires were distributed with an aim to analyze atleast 385 as suggested by Glenn (1992).

$$n_0 = \frac{Z^2 pq}{e^2} = \frac{(1.96)^2 (.5) (.5)}{(.05)^2} = 385 Students$$

However, a total of 396 data found to be analyzable which represents 82% after screening and cleaning processes are accomplished. Data were obtained through social media sites and face-to-face distribution. The 396 respondents were surveyed for the study indicated a diverse demographic profile where male responded higher than female (53% are male and 47% are female). Most of the respondents from Klang Valley are aged between 22 and 25 indicates 48.5%, followed by the age between 26 and 30 (33.58%). Respondents then were asked about their educational qualifications, found 58.83% of them were pursuing undergraduate degree. Lastly, 47.72% responses were from Malay ethnic group, followed by Chinese (25.76%), Indian (17.93%) and others (8.59%).

Result Analysis

Exploratory Factor Analysis (EFA)

Exploratory Factor Analysis (EFA) is a statistical method for evaluating the association between variables in the dataset. Factor analysis is generally performed to reduce data. Large set of variables are taken to explore a way of how the data reduction can be useful to explain smaller set of factor or component. The Kaiser-Meyer-Olkin (KMO) test is usually performed for research purposes. The purpose of this study is to determine the degree of unidimensionality of the scales in the data collected. Total of 15-items of the four variables were factor analyzed. The Kaiser-Meyer-Olkin (KMO) and Bartlett test were carried out in order to validate adequate data for the purpose of factor analysis and the results are shown in Table 1.

Table 1:KMO and Bartlett's TestKaiser-Meyer-Olkin Measure of Sampling Adequacy..835Bartlett's Test of SphericityApprox. Chi-Square1650.633df105

Sig.

.000

For data purification, exploratory factor analysis was conducted using the key axis factoring and varimax rotation method. The varimax rotation technique is used to produce a matrix containing coefficients or loading factors representing the association of factors and variables. KMO and Bartlett indicate the higher values obtain from KMO and Bartlett is suitable enough for factor analysis. Four-factor solutions with a KMO statistic of 0.835 indicates that the sample size was adequate while further Bartlett's test of Sphericity was found significant at .000 and a chi-square statistic of 1650.633 with 105 degrees of freedom. The total variance in the five factors explained was 72.417. The Eigen value is commonly used to calculate the number of variables, and a cut-off value of 1 is typically used to identify factors that are dependent on eigenvalues. The eigenvalues for the four factors were 5.087, 2.263, 1.844 and 1.667 respectively.

The findings suggest that the data were suitable for factor analysis. Things were aggregated resulting in the discovery of four main factors. For the purpose of the report, these four factors are identified with new names, which are Factor 1 as Technology self-efficacy, Factor 2 as Perceived Ease of User, Factor 3 as Perceived Usefulness and, finally, Factor 4 as Satisfaction. The study then used Cronbach's alpha for further reliability test calculation.

Using SPSS, the alpha value of 0.892 for Technology self-efficacy (4 items), 0.809 for Perceived Usefulness (4 items), 0.864 for Perceived Ease of Use (4 items) and 0.854 for student satisfaction (3 items) were calculated which explains that all the items in the scale used to measure the variables taking part are reliable. Cronbach's alpha should have 0.7 or higher than that are recommended by Pallant (2007).

Confirmatory Factor Analysis (CFA)

A multivariate statistical method is called Confirmatory Factor Analysis (CFA) which has been used to determine how well the number of constructrs are represented by measured variables. The proposed framework for measurements (Figure 2) was therefore tested using AMOS.



Figure 2: Measurement Model

CFA was performed to evaluate the factor structure that consists of a 15-item scale using AMOS. The association of four latent constructs has been presumed. Based on the modification indices presented by AMOS, no indicatores were eliminated from the measurement model as loadings of each items were found to be 0.7 and above. Hair et al. (2010) signified any loading value below 0.5 will be considered insignificant. To ensure a good model fit, Table 1 displays nine good fit model indices. According to Hair et al. (2010), the value of GFI and CFI are suggested to be more than 0.95 and RMSEA should be smaller than 0.05 in order to clarify well fit. For x^2 /df, below 3 was considered good value and acceptable. TLI should at least be 0.90 or above than that value.

Table 2:								
Model Fit								
χ2	df	^{x2} /df	GFI	RMSEA	NFI	CFI	IFI	TLI
104.480	84	1.244	0.942	0.034	0.938	0.987	0.987	0.984

The 15-item scale shows the good model fit and acceptable as the minimum thresholds concerted.

In order to achieve good composite reliability, it was suggested that the minimum agreed value be 0.7 as recommended by Chin (1988). In addition, the average extracted variance (AVE) value should be at least 0.5 or even more, as suggested by Hair et al. (2006). Present study was calculated AVE to prove good indication. Standardized loadings of each item were used to calculate AVE where, sum of the squared standardized loadings were divided by the number of indicators. AVE found to be ranging from 0.516 to 0.681 which was greater than a minimum threshold of 0.5. Study also explained CR value ranging from 0.810 to 0.965 that exceeded the ideal threshold value of 0.7. Therefore, in order to maintain discriminatory validity, it is important to prove that the measures which should not be related are not actually related (Hair et al., 2013). Discriminating validity can also be measured by the square root of AVE (Fornell & Larcker, 1981). Table 3 shows the high discriminant validity as it demonstrates that the square AVE of each factor exceeds all its correlations with the other factors.

$$AVE = \frac{\sum \lambda_i^2}{\sum \lambda_i^2 + \sum_i \text{Var}(\varepsilon_i)}$$
$$CR = \frac{(\sum \lambda_i)^2}{(\sum \lambda_i)^2 + \sum_i \text{Var}(\varepsilon_i)}$$

Table 3: Confirmation Factor Analysis (CFA) Report

Variables	AVE	CR	TSE	PU	PEOU	SAT
Technology Self-Efficacy (TSE)	0.679	0.965	0.824			
Perceived Usefulness (PU)	0.516	0.810	0.430	0.718		
Perceived Ease of Use (PEOU)	0.617	0.865	0.351	0.288	0.785	
Satisfaction (SAT)	0.681	0.863	0.311	0.189	0.336	0.825

Note: AVE= Average variance extracted, CR= Composite reliability

Structural Model

The six (6) hypotheses were examined in this research to look at the causal pathways. The results of hypothesis testing are described in Table 4 on the basis of the hypothesized structural model. Among the six tested hypotheses, four are statistically significant (p < 0.05) and the remaining two failed to be significant. R² value was evaluated. Technology self-efficacy accounted for 12.4% of variance (R²= 0.124) in explaining perceived ease of use and 20.6% of variance (R²= 0.206) in explaining perceived usefulness. Meanwhile, Technology self-efficacy, perceived ease of use and perceived usefulness were accounted for 15.6% of the variance (R²= 0.156) in explaining student satisfaction towards distance learning apps.

The results of hypotheses H1 signify that a positive association between technology selfefficacy and student satisfaction is supported based on β = 0.211; SE = 0.078; CR = 2.491. Similarly H2 and H3 were supported based on the β value which shows 0.375 for H2 and 0.351 for H3 respectively. H2 also shows the SE= 0.060; CR= 4.371 and significant P value which indicated that technology self-efficacy have positive and significant influence on perceived usefulness. Similar result can be seen for H3 where SE=0.057; CR= 4.424 thus, H3 was also supported by explaining that there is a positive relationship between technology self-efficacy and perceived ease of use. On the other hand, H4 shows that there was no significant

relationship is associated between perceived ease of use and perceived usefulness. With the value of β = 0.156; SE= 0.080 and CR= 1.909 and p value was higher than 0.05 the hypothesis (H4) was not supported. H5 found to be significant based on the p value (p<0.05) which indicates a positive relationship between perceived ease of use and satisfaction. H6 which express that no positive relationship is associated between perceived usefulness and student satisfaction. Thus, result of H6 is rejected based on β = 0.025; SE=0.111 and CR=0.289.



Figure 3: Structural Model

Discussion

The findings of this study reveal that technology self-efficacy is an important factor among university students during COVID-19 pandemic. In terms of distance learning satisfaction student have perceived how well he/she can afford in paying attention to online learning since the number of educational institutions are shifting from face-to-face instruction to online based learning. Current study has successfully proved the effectiveness of technology and knowledge regarding to use it (H1 was supported). Although Islam (2020) failed to signifiv the postive relationship between technology self-efficacy and satisfaction, current study found that without technologial sense distance learning is not satisfactory. Similarly, perceived technological sense among students are necessary for understanding the effectiveness of distance learning. Current study found the positive association between technology self-efficacy and perceived usefulness. Thus, H2 is in line with the prior study of Chen (2012). Consequently, if the technology is perceived by students as useful, simple and easy, they will eager to accept it (Abdullah & Ward, 2016). The study then investigated the relationship between PEOU and PU. Although online learning in previous studies have found to be convenient and easy, current study failed to explain the significant relationship with Perceived usefulness. Although several studies have interpreted the technology based learning system to be useful if students perceived that it is clear and understandable (Lee, 2006; Shah and Attiq, 2016), current reseach found that during the COVID-19 outbreak students satisfaction towards distance learning systems are based on how easy and accessible the designed system rather than its effectiveness. Hence, perceived ease of technology use

for distance learning doesn't influencing its usefulness and therefore ended up by rejecting H4.

Study findings suggest that perceived ease of use have huge effects on student satisfaction (H5 was supported). From the findings it can be concluded that easy to use distance learning programs are of vital importance for student satisfaction during COVID-19. Result of this study is consistent with the prior research of Islam (2020). To validate the extended TAM model both satisfaction and self-efficacy were meta analyzed where perceived ease of use found to be important factor in determining satisfaction among students. In addition, students have been asked to do social distance, saying that all educational institutions should encourage online learning and teachers should be encouraged to use the systems in a much easier way. On the contrary, perceived usefulness of distance learning apps found to have no significant impact on student satisfaction. Student perceived that learning through technology is satisfactory if the system is easy to operate rather than its evvectiveness. This is because those who were not very proficient in online learning due to the corona virus, they want this system to make it easier for them. Thus H6 is not supported.

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Н	Relationship	SRW (β)	S.E.	C.R.	p- value	Decision
H1	SAT←TSE	0.211	0.078	2.491	0.013	Supported
H2	PU←TSE	0.375	0.060	4.371	***	Supported
H3	PEOU←TSE	0.351	0.057	4.424	* * *	Supported
H4	PU←PEOU	0.156	0.080	1.909	0.056	Not Supported
H5	SAT←PEOU	0.255	0.105	3.110	0.002	Supported
H6	SAT←PU	0.025	0.111	0.289	0.772	Not Supported

Table 4: Hypotheses Testing

Note: *p<0.05; **p<0.01; ***p<0.001

Contributions and Implications

This study provides quantitative insights into the latest trend and an understanding of students' distance learning satisfaction during the novel Coronavirus COVID-19. In addition, the results of this study suggest the reliability and validity of the Technology Acceptance Model in the assessment of student satisfaction and its relationship with the preceding variables. To support the theory, current study has employed perceived usefulness and perceived ease of use to direct the relationship with satisfaction. In TAM model perceived ease of use and usefulness was hypothesized as important precursors to predict the behavioral intention. Current work is also an important addition to TAM by proposing a model to resolve an antecedents of perceived usefulness and perceived ease of use. In the online learning environment, technology knowledge is mandatory. On account of this, technology self-efficacy was designated as an external factor of TAM model.

The results of this study give Malaysian education sector a few managerial implications. Firstly, the findings of this study will allow the educational institution 's authority to better understand the best strategy for the education sector to be enforced. In addition, app developers can take important steps to make their apps easier for students to use. Because the findings of this study reveal that Satisfactory attitude of students depends on the ease of use of these distance learning apps because the easier the apps are to use, the more motivated the students will be to use them. Currently, a number of educational institutions

in Malaysia have shifted from face-to-face classes to online classes for Covid-19, as a result of which all kinds of activities are being performed online. Online classes have not been very well conducted in Malaysia before, so students can be a little hesitant about the proper use of this technology in education. For this, developers should consider how to make apps easier for students.

Limitations and Future Research

This study has few limitations. First, due to time constraints, researchers have only given much priority to the Klang Valley area in Malaysia, where university students living in that particular area are preferred. Since this research has adopted a convenient sampling technique and selected the Klang Valley Zone, there are many potential candidates for data collection who are outside the Klang Valley Area and have been excluded from the survey. Second, even though the movement control order (MCO) was lifted by imposing Standard Operating Procedure (SOP) in Malaysia due to the numbers of Covid-19 cases were dropped, some students could not participate in the survey due to security reasons, which resulted in the exclusion of a significant portion of the participants. However, it is further recommended that future study could cover throughout the Malaysia including Sabah and Sarawak as the numbers of students in that particular area are not low. Convenient sampling method has widely being applied by researchers to generalize the results but also it has been criticized by many researchers as well. Future research could adopt simple random sampling method so the likelihood for selection for each member of the subset is equal.

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