



Research Article

The Mediating Role of Student Satisfaction in the Relationship between Determinants of Online Student Satisfaction and Student Commitment

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Abstract

The present study uses a proposed model to explore satisfaction and commitment of students in an online learning environment. First, it presents the key determinants of student satisfaction and their impact on student commitment. Second, it examines the significance impact of each of the three key determinants namely course structure, online tutorials flexibility and technology quality on both student satisfaction and commitment. A sample of 410 students enrolled in an Egyptian higher education institution pursued an online questionnaire through Survey Monkey. This study used structured equation model approach for data analysis that was gathered online. However, the only determinant namely course structure had insignificant effect on both student satisfaction and student commitment. Specifically, both key determinants namely online tutorial flexibility and technology quality had a significant influence on student satisfaction as well as student commitment. In addition, the mediating role of student satisfaction was tested and had positive influence between the key determinants and student commitment except with course structure. The study highlights the issue of commitment and satisfaction of students in an online learning environment with respect to some determinants. The context was limited to one higher education institution in an online learning environment. Further research is needed to investigate the key determinants in other higher educational institutions in Egypt and could be extended to other countries if possible. This study will be useful for academics and educators interested in online research behavioral issues of students.

Keywords: satisfaction, commitment, online learning, course structure, online tutorial flexibility, technology quality

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Introduction

Over the previous years, higher education institutions were offering courses online for students as one of their academic plan's components. Online courses are connecting a variety of tools such as social networking and curriculum available online through different e-learning platforms such as Moodle, Blackboard, etc. The active involvement of many students in these courses who participate individually according to their achieving goals, previous knowledge, and skills, is a major factor for success. (McAuley et al., 2010). Education is one of the most dominant means of providing social equality as well as an accomplishing experience of an individual's life. The basic tools require computer screens connected to the Internet so lectures could be delivered, the course content is visible for the students to read, audio devices are required to deliver the course content as well as interactions and discussions via video output. Many academic institutions used Zoom as their platform to deliver all course from video and audio tools as well as Moodle for the whole course content and syllabus from power points, pdf files, assignments, and quizzes.

E-learning is defined as the learning tool delivered via computers for the intention to develop education (Mayer, 2003) that has expanded quickly over last years with the upgrades in technology and the integration of IT with educational curricula. (Smart and Cappel, 2006). E-learning is essential to develop good calibers, well educated, as well as the rapid need of society for long term learning that is delivered in an appropriate form. Learning online can take many forms from fully online, blended or web-assisted, no matter what the delivery technique of learning, however different tools are available at the students' and faculty members' hands. (Alshehri, 2017). In fact, online learning is a tool that is globally used for learning with standards and it is not required from the learner or the student to attend the class physically and is suitable for students who are interested in studying while they have other commitments. (El-Ebiary et al., 2016).

E-learning now is not a strategy to overcome just learning from distance, it is now a way to adapt to learning as per personal schedule especially

after the pandemic of COVID-19 2020. In this study, students used Moodle that allowed them to interact with the lecturers using audio and video, submit assignments and quizzes, upload and/or download files, have feedback on their grades and have access to the content anywhere anytime. The development of such platforms or applications made the online learning a more interactive way for students to learn at their preference. In recent years, interventions using technologies based on the Internet have allowed electronic learning as a major element in education (Aşkar and Halici, 2004). Online learning can provide academic institutions with a reduced cost, and adaptable tool to grow globally (Casey, 2008).

Students enjoyed this way of online delivery and sometimes it seemed to be more effective than face to face teaching. While students have an online meeting with their instructors, this screen sharing approach may help the students feel less exposed when asking for help compared to a physical class within groups. The students may prefer a quicker way to ask for help such as posting their questions on the online chat. This way of asking questions may allow the student to feel more comfortable than raising up their hands in front of their colleagues in a computer lab. Similar recent studies have reported (Davison, 2020) that student interaction was motivated through the use of online chat functionality. (Dwivedi et al., 2020).

In light of the dramatic increase in the use of online courses, in higher education especially after the pandemic of COVID-19, the current study deals with identifying key determinants of student satisfaction in an online learning environment namely online tutorials flexibility, online tutorials quality and technology quality. It also investigates the impact of the student satisfaction on their commitment towards the institution and the learning process. Several studies tackled student satisfaction and its key determinants but one of the recent studies investigated factors that were relevant to the online learning environment students have experienced through the pandemic in Egypt. The study by Harsasi and Sutawijaya (2018) included four variables namely course structure, online tutorials flexibility, online tutorials quality and technology quality but online tutorials quality

had less impact on student satisfaction. Therefore, the proposed model investigates three variables namely course structure, online tutorials flexibility and technology quality according to the pandemic period in Egypt.

This study sheds light on some of the determinants of students' satisfaction and the impact of students' satisfaction on their commitment online. This is to know which determinants are the most effective on the students' satisfaction in the online learning process and to explore the influence of their online satisfaction on their commitment towards their institution. Technology becomes a crucial part in our life. E-learning, Internet and computers have been used extremely in the learning process during COVID-19 (Khan and Raad, 2020). This study may have a marked impact on online course satisfaction and commitment. Furthermore, this study may affect not only academic courses but have paved the way for academic institutions and professional training centers to measure the behavior of students after COVID-19. Instructors and researchers can use this study and the preliminary model for the benefit of their research and instruction specially to understand and improve the behavior of students through the online learning process.

Literature Review

There are a number of studies that have investigated student satisfaction and commitment in relation to e-learning. A study found that the less student satisfaction the more the failure of any e-learning implementation, and there are many determinants that affect student satisfaction in e-learning. Moreover, Hermans et al. (2009) suggested that student satisfaction has a vital role to endorse higher education successfully. These authors tested the relationships among different factors that influence student satisfaction in an online learning environments. (Alshehri, 2017)

A definition for satisfaction and commitment needs to be clarified. Student satisfaction is an indicator of whether learners or students are satisfied with their learning experience. (Li et al., 2016). In the academic context of higher education, student satisfaction plays an important role in universities' success (Firdaus, 2006), and the concept of satisfaction has been extended to evaluations of higher education services. Elliott and Healy (2001) state that the

concept of students' satisfaction is a short-term behavior and a result of students' experiences with their educational services provided by their institution.

Previous research has found that organizational commitment which is defined as a person's commitment to the organization they work in is related to their career commitment (Womack, 2016), which means that a person's commitment to their organization is an indicator of their commitment to his or her career. The concept of organizational commitment is altered to state the commitment that the student feels towards his or her institution. (Womack, 2016)

Commitment can be viewed as future-oriented self-regulation where one sets the goals based on their identity and then behaves in a way to achieve the goals bounded by their identity. (Human-Vogel and Vogel, 2015).

A tool was developed to measure employee commitment based on a three-component model of affective, continuance and normative commitment. Affective commitment refers to the attachment and involvement to the organization, Continuance commitment refers to the costs assessed of leaving the organization and Normative commitment refers to the feeling of supporting the organization and remaining in it. (Wilson et al., 2016). The aim of a recent study was to explore factors that have an effect on student satisfaction in online learning environment based on previous research by (Sun et al., 2008) and (Eom et al., 2006).

The impact of course structure on student satisfaction

Course structure is viewed as an important variable that supports the success of online learning. According to Moore (1991), the course structure "expresses the rigidity or flexibility of the program's educational objectives, teaching strategies, and evaluation methods" and the course structure describes "the extent to which an education program can accommodate or be responsive to each learner's individual needs."

Course structure has two parts which are course objectives and course infrastructure. Course objectives are concerned with the course curriculum that includes the topics to be learned, the assignments to be completed, the class participation online, group projects and so on. Those parts affect the student satisfaction level

and their learning outcomes. According to Eom et al. (2006), it was stated that course structure will be correlated to student satisfaction and perceived learning outcomes when the course material is organized in a logical sequence and that a clarified course objectives will lead to high student satisfaction levels. (Eom et al., 2006)

Hypothesis 1: Course structure has a significant impact on student satisfaction.

The impact of online tutorials flexibility on student satisfaction

Satisfaction and participation of students are facilitated because of the flexibility in time, location and methods found in online learning courses. (Arbaugh, 2002) (Arbaugh, 2000) (Berger, 1999). Moreover, the exclusion of any traditional classroom environment enables more interaction that promotes cooperative learning. Students can communicate online anytime anywhere with no limits. In addition, the virtuality available online reduces the discomfort of face-to-face communication that is available in traditional classrooms. Students have the freedom to express their opinions and ask questions through discussion groups without restraint. "The definition of e-Learning course flexibility is learners' perception of the efficient and effects of adopting e-Learning in their working, learning, and commuting hours". (Sun et al., 2008)

Hypothesis 2: Online tutorials flexibility has a significant impact on student satisfaction.

The impact of technology quality on student satisfaction

Several researchers suggest that the technology quality and Internet quality affect satisfaction in e-Learning (Piccoli et al., 2001) (Webster, 1997). Students will be willing to adopt an e-learning software or tool with few barriers and their satisfaction will be improved. (Amoroso and Cheney, 1991). Online learning may involve learning and discussion using tools such as video conferencing. Therefore, quality of the technology and the Internet are important factors in e-Learning (Piccoli et al., 2001) (Webster, 1997). In addition, a research study conducted by Webster (1997) studied learning effects on the technology applied in a distance learning of 247 students and found that quality and reliability of technology influence learning

effects. "The definition of technology quality is the learners' perceived quality of IT applied in e-Learning (such as microphones, earphones, electronic blackboards, and so on)". (Sun et al., 2008)

Hypothesis 3: Technology quality has a significant impact on student satisfaction.

The impact of student satisfaction on commitment

In an academic context, the knowledge about teachers' attitudes needs more attention as the universities' performance depends on their staff which must be satisfaction and commitment (Tsui and Cheng, 1999). In the same manner, students are other academic entities and their satisfaction and commitment require more attention. An academic employee will feel more satisfied, more committed to his or her academic institution and will develop the output of their universities (Sami et al., 2012). Some research studies support the contention that satisfaction predicts commitment and that satisfaction is positively related with affective and normative commitment, but not with continuance commitment. (Bashir and Ganai, 2019)

Hypothesis 4: Student satisfaction has a significant impact on student commitment.

Research Hypotheses

The main aim of this study is to explore the impact of each of those variables on student satisfaction in an online learning environment and the mediating effect of student satisfaction between independent variables and student commitment.

Based on the previous study, this research adopted three variables that have influence on student satisfaction online. These variables are course structure, online tutorial flexibility and technology quality. To achieve the research objectives, the hypotheses were developed as follows:

H1: Course structure has a significant impact on student satisfaction.

H2: Online tutorials flexibility has a significant impact on student satisfaction.

H3: Technology quality has a significant impact on student satisfaction.

H4: Student satisfaction has a significant impact on student commitment.

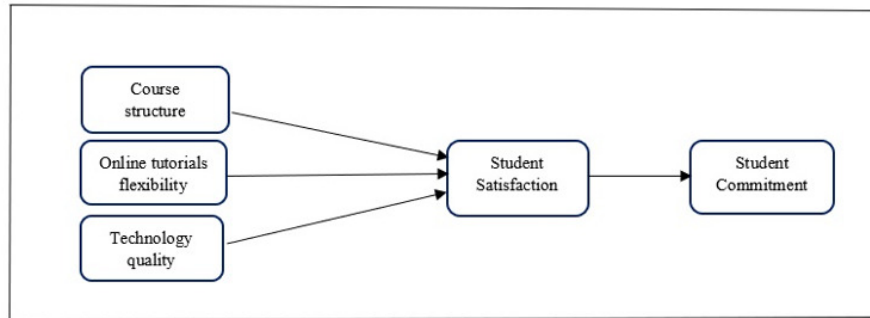


Figure 1: The proposed research model

Methodology

This research was based on the use of three independent variables (course structure, online tutorials flexibility and technology quality) and one mediating variable (student satisfaction). In addition, the dependent variable student commitment was used based on another previous study by Wilson et al. (2016).

The observed data were collected using a survey methodology. The target population were students from higher education institutions, and due to the pandemic of COVID-19, the sample size was 410 students inside one of the top accredited universities in Egypt. Participants of the study were students from different colleges inside the university. The students are those who pursued Second Semester 2020/2021 and who took the experiment of online courses through the pandemic of COVID-19.

A questionnaire was distributed to the students online through a link on

SurveyMonkey. The responses to the questionnaire were 410 students from different colleges such as College of Management and Technology, College of International Transport & Logistics, College of Language & Communication, College of Computing & Information Technology and College of Engineering & Technology.

The survey instrument included the independent variables and the mediating variable were adopted from Harsasi and Sutawijaya (2018) while the dependent variable was adopted from Wilson et al. (2016). Each variable's indicators are measured using 5 points Likert scale with the following scale:

- 1 = Strongly disagree
- 2 = Disagree
- 3 = Neither agree nor disagree
- 4 = Agree
- 5 = Strongly agree

Each variable has their own indicators as shown in Table 1.

Table 1: Indicators of each variable

Variable	Indicators	Code
Course Structure	1. Course material is presented in an organized structure	CS1
	2. The learning objectives in the online tutorial has been conveyed properly	CS2
	3. The material in the online tutorial has been understandable and arranged in a logical sequence	CS3
	4. The structure of the syllabus in the online tutorial already covers all the material I need to learn in one course	CS4
Online tutorial flexibility	1. Learning through online tutorial gave me the flexibility to adjust my learning time	OTF1
	2. Learning through online tutorial benefit me	OTF2
	3. Learning through online tutorial made me have the flexibility to divide the time between learning activities and other jobs	OTF3
	4. Learning through online tutorial has no disadvantage	OTF4
	5. Learning through online tutorial lets me manage my time more effectively	OTF5
	6. Learning through online tutorial makes me save time rather than having to attend class	OTF6
	7. Learning through online tutorial saved me from not missing the lecture as I can learn the material at any time	OTF7
Technology quality	1. I can access online learning anywhere	TQ1
	2. I do not experience any problems when learning online	TQ2
	3. I do not encounter any difficulty in responding to the discussion	TQ3
	4. I do not see any difficulty when uploading task, I feel that technology for online learning is:	TQ4
	Easy to use	TQ5
	Have useful functions	TQ6
	Very helpful for learning materials	TQ7
	Facilitate communication with tutors or other students	TQ8
Student Satisfaction	1. I am satisfied with the whole system of online learning	OSS1
	2. Overall, online learning system is already well	OSS2
	3. Overall, online learning has been successful	OSS3
	4. Learning through online learning system enable me to learn independently	OSS4
	5. I will keep learning through the online learning system in the future	OSS5
Student Commitment	1. I am happy being a student of AASTMT	SC1
	2. I enjoy discussing about AASTMT with the people outside it	SC2
	3. I really feel as if AASTMT's problems are mine	SC3
	4. I am deeply attached to AASTMT	SC4
	5. I am part of the family of AASTMT	SC5
	6. I feel emotionally attached to AASTMT	SC6
	7. AASMT has a great personal meaning for me	SC7
	8. I am afraid of what might happen if I quit my studies in AASMT without having another one lined up	SC8
	9. It would be very hard for me to leave my studies in AASMT right now, even if I wanted to	SC9
	10. Too much in my life would be disrupted if I decided to leave my studies in AASTMT now	SC10
	11. It would be too costly for me to leave AASTMT now	SC11
	12. Right now, staying on my studies in AASMT is a matter of necessity	SC12
	13. I think that people these days rarely move from a university to another	SC13
	14. I believe that a person must always be loyal to his or her AASMT	SC14
	15. Jumping from AASMT to another seems unethical to me	SC15
	16. One of the major reasons I continue to study in AASTMT is that I feel a sense of moral obligation to remain	SC16
	17. Even if I got another offer of a better university elsewhere, I would feel it is right to stay in AASTMT	SC17
	18. Things were better in the days when people stayed in one institution for most of their career	SC18

Results and Discussion

This section presents the data analysis part of the study. The analysis of this paper was done using the statistical package for social sciences (SPSS V26) for basic descriptive statistics, and (SmartPLS 3.2.7) for SEM-PLS modeling. It is divided into four sections respectively: Data preparation, measurement model for reliability and validity, several descriptive statistics and bivariate correlations are constructed and finally the structural model for hypothesis testing and mediation analysis.

Data preparation

This examination is important in any quantitative research and specifically when using SEM for data analysis (Hair Jr et al., 2017). The issue of missing data was inspected and found that some indicators have missing percent greater than 5%, so, according to Hair Jr et al. (2017) the best scenario is the case-wise deletion. Considering outliers, there are no outliers detected in our dataset.

The data distribution is not an issue, since the SEM-PLS is a non-parametric tool that does not assume normal data ((Hair Jr et al., 2017)

(Garson, 2016). Structural equation modeling is an analysis approach that tests both the measurement model and path that helps to develop more realistic assumptions (Abdi, 2010) (Hair Jr et al., 2017). Hence, this study focuses on examining the prediction of the dependent variable, the mediation analysis, and the effect of independent variables, which make PLS method the most appropriate in this study.

Assessing the Measurement Model

The measurement models, which are also known as the outer models, describe the relationships between the constructs and their items. The assessment of the reflective measurement models in PLS-SEM requires evaluating the internal consistency reliability, convergent validity, and discriminant validity (Hair et al., 2011) (Nachtigall et al., 2003) (Xiong et al., 2015) (Garson, 2016). The internal consistency reliability examines whether all of the indicators associated with a construct are actually measuring it (Pallant and Manual, 2010). There are different ways to measure internal consistency. Cronbach's alpha is a statistical measure that is the most used for this purpose. The accepted value of Cronbach's alpha is 0.7; all values of Cronbach's alpha in the table (2) were above 0.7.

Table 2: Results of the Measurement Model

Latent Variable	Indicator	Loadings	AVE	CR	rho_A	Cronbach's Alpha	Discriminant Validity
Cut-off Point		> 0.4	> 0.5	> 0.7	> 0.7	> 0.7	HTMT ratio Less Than 0.9
Course Structure	CS1	0.829					
	CS2	0.912	0.792	0.938	0.921	0.912	Yes
	CS3	0.916					
	CS4	0.9					
Online Tutorials Flexibility	OTF1	0.849					
	OTF2	0.846					
	OTF3	0.875					
	OTF4	0.688	0.683	0.937	0.923	0.921	Yes
	OTF5	0.872					
	OTF6	0.818					
	OTF7	0.821					
Technology Quality	TQ1	0.686					
	TQ2	0.798	0.664	0.94	0.932	0.927	Yes
	TQ3	0.829					

	TQ4	0.843					
	TQ5	0.797					
	TQ6	0.84					
	TQ7	0.891					
	TQ8	0.819					
Student Satisfaction	OSS1	0.919					
	OSS2	0.931					
	OSS3	0.929	0.783	0.947	0.937	0.93	Yes
	OSS4	0.774					
	OSS5	0.862					
Student Commitment	SC1	0.784					
	SC2	0.82					
	SC3	0.822					
	SC4	0.851					
	SC5	0.825					
	SC6	0.865					
	SC7	0.882					
	SC8	0.808					
	SC9	0.77					
	SC10	0.782	0.637	0.969	0.967	0.966	Yes
	SC11	0.737					
	SC12	0.688					
	SC13	0.63					
	SC14	0.856					
	SC15	0.796					
	SC16	0.82					
	SC17	0.823					
	SC18	0.761					

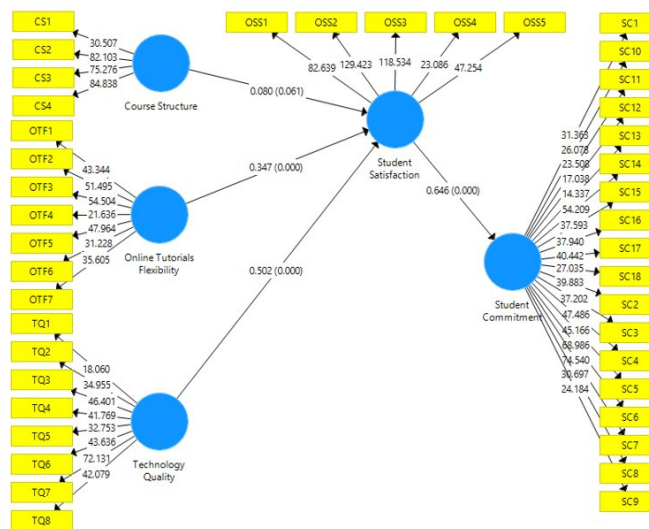


Figure 2: Path Coefficients with Corresponding P-values

The AVE is a standard measure used to establish convergent validity. All of the constructs in Table (2) have AVE scores higher than 0.50. After establishing the convergent validity, it is time to examine the discriminant validity. Discriminant validity examines how much a construct differs from other constructs.

Descriptive Statistics and Multiple Correlations

Table (3) shows the descriptive statistics of the variables. It can be shown that course structure has mean and standard deviation as ($M = 3.67, SD = 0.943$) with direct medium correlation to both online student satisfaction ($r = .683, P < 0.001$), and student commitment ($r = .611, P < 0.001$).

Table 3: Descriptive Statistics and Multiple Correlations

	CS	OTF	TQ	OSS	SC	
Multiple Correlations	CS	1	.718***	.712***	.683***	.611***
	OTF		1	.806***	.810***	.579***
	TQ			1	.837***	.688***
	OSS				1	.645***
	SC					1
Descriptive Statistics	Mean	3.6732	3.6504	3.7299	3.5813	3.7695
	Std. Deviation	0.94271	0.97918	0.85533	1.00841	0.88209
	Skewness	-0.746	-0.775	-0.662	-0.742	-0.718
	Kurtosis	0.271	-0.006	0.025	-0.088	0.152

*** $P < 0.001$

Online tutorial flexibility has mean and standard deviation as ($M = 3.65, SD = 0.979$), with direct strong correlation to online student satisfaction ($r = .810, P < 0.001$), and direct medium correlation to student commitment ($r = .579, P < 0.001$). Technology quality has mean and standard deviation as ($M = 3.73, SD = 0.855$), with direct strong correlation to online student satisfaction ($r = .837, P < 0.001$), and direct medium correlation to student commitment ($r = .688, P < 0.001$). Online student satisfaction with mean and standard deviation ($M = 3.58, SD = 1.01$), and student

commitment with mean and standard deviation ($M = 3.77, SD = 0.882$) correlate to each other with direct medium correlation as ($r = .645, P < 0.001$).

Assessing the Structural Model

Researchers provided guidelines for evaluating and reporting the structural model, including path coefficients, collinearity, coefficient of determination (R^2), effect size (f^2), predictive relevance (Q^2), and goodness of fit criteria; table (4) summarises those criteria.

Table 4: Criteria of Structural Model Assessment

Criteria	Guidelines	Reference
Path coefficients	Significance: $p \leq 0.05$	(Hair Jr et al., 2017)
Collinearity	VIF < 5	(Hair Jr et al., 2017)
Coefficient of determination (R^2)	Weak effect: $R^2 = 0.19$ Moderate effect: $R^2 = 0.33$ High effect: $R^2 = 0.67$	(Chin et al., 1998)
Effect Size (f^2)	f^2 between 0.02-0.14, small; f^2 between 0.15-0.34, moderate; $f^2 \geq 0.35$, High.	(Chen et al., 1998)
Cross-validated redundancy (Q^2)	Predictive Relevance Using blindfolding $Q^2 > 0$	(Chin et al., 1998)
Goodness of Fit (GoF)	GoF less than 0.1, no fit; GoF between 0.1 to 0.25, small; GoF between 0.25 to 0.36, medium; GoF between 0.25 to 0.36, large.	(Wetzels et al., 2009)

Path coefficients refer to the estimates of the relationships between the model's constructs. When assessing the PLS path, studies should report path coefficients beside the significance level, t-value,

and p-value. According to Hair Jr et al. (2017), every significant coefficient eventually depends on the standard error, which is usually obtained by using bootstrapping. Moreover, the strength of path

coefficients can be measured through direct and indirect effects. We are interested in some direct and indirect effects to test the hypothesis underlying this research.

The hypothesis testing has been done to understand the signs, size, and statistical significance of the estimated path coefficients between the constructs. Higher path coefficients suggest stronger effects between the predictor and predicted variables. The significance of the supposed relationships has been established by measuring the significance of the p-values for each path with threshold equalling $p < 0.05$. The p-values and inference of hypotheses, as well as the confidence level for each estimate, are shown in Table 4. The findings of this study reveal that all hypotheses are accepted except H1 and H5.

The results show that online tutorials flexibility yielded a significant effect on student satisfaction since ($\beta = 0.347, t = 4.798, P < 0.001, 95\% CI \text{ for } \beta = [0.208, 0.493]$), so the 2nd

hypothesis is accepted. Technology quality yielded a significant effect on student satisfaction since ($\beta = 0.502, t = 6.858, P < 0.001, 95\% CI \text{ for } \beta = [0.351, 0.637]$), so the 3rd hypothesis is accepted. Student satisfaction yielded a significant effect on student commitment since ($\beta = 0.646, t = 15.87, P < 0.001, 95\% CI \text{ for } \beta = [0.561, 0.724]$), so the 4th hypothesis is accepted. The mediation analyses showed that student satisfaction mediates the relationship between online tutorials flexibility and student commitment through the indirect effect ($\beta = 0.224, t = 4.759, P < 0.001, 95\% CI \text{ for } \beta = [0.135, 0.32]$), so the 6th hypothesis is accepted. Student satisfaction also mediates the relationship between technology quality and student commitment through the indirect effect ($\beta = 0.325, t = 6.009, P < 0.001, 95\% CI \text{ for } \beta = [0.216, 0.429]$), so the 7th hypothesis is accepted. The remaining hypotheses are not supported.

Table 5: Hypothesis Testing

	??	Mean	Std. Deviation	t-value	P-value	95% CI		Decision
						LL	LL	
Direct Effects								
H1: Course Structure -> Student Satisfaction	0.08	0.081	0.043	1.872	0.061	-0.001	0.164	Not Supported
H2: Online Tutorials Flexibility -> Student Satisfaction	0.347	0.348	0.072	4.798	0.000***	0.208	0.493	Supported
H3: Technology Quality -> Student Satisfaction	0.502	0.501	0.073	6.858	0.000***	0.351	0.637	Supported
H4: Student Satisfaction -> Student Commitment	0.646	0.648	0.041	15.87	0.000***	0.561	0.724	Supported
Indirect Effects								
H5: Course Structure -> Student Satisfaction -> Student Commitment	0.052	0.053	0.028	1.833	0.067	-0.001	0.11	Not Supported
H6: Online Tutorials Flexibility -> Student Satisfaction -> Student Commitment	0.224	0.225	0.047	4.759	0.000***	0.135	0.32	Supported
H7: Technology Quality -> Student Satisfaction -> Student Commitment	0.325	0.325	0.054	6.009	0.000***	0.216	0.429	Supported

Table 6: Structural model assessment measure

	Student Satisfaction		Student Commitment	
	Effect size	VIF	Effect size	VIF
<i>Course Structure</i>	0.011 (No Effect)	2.344	-	-
<i>Online Tutorials Flexibility</i>	0.154 (Moderate)	3.224	-	-
<i>Technology Quality</i>	0.325 (Moderate)	3.2	-	-
<i>Student Satisfaction</i>	-	-	0.717 (High)	1
<i>Student Commitment</i>	-	-	-	-
R Square	0.758 (Strong)		0.417 (Moderate)	
R Square Adjusted	0.755 (Strong)		0.416 (Moderate)	
Q²	0.584		0.26	

Coefficient of determination (R^2) refers to the effect of independent variables on the latent dependent variables (Hair et al., 2011), which is one of the quality measures of the structural model (Hair et al., 2014). Researchers have used a different cut-off of R^2 value. For example, Hair et al. (2011) in marketing research described that R^2 values of 0.25, 0.50, or 0.75 are low, moderate, or high, respectively. In business research, Chin et al. (1998) suggested that R^2 with 0.19, 0.33, or 0.67 are low, moderate, or high, respectively. The results of R Square are reported in table (6), the R-Square of student satisfaction equals 76% which means that about 76% of the variations in student satisfaction are explained by the variations in the selected independent latent variables. Additionally, the R-Square of student commitment equals 42% which means that about 42% of the variations in student commitment are explained by the other latent variables.

The f^2 effect size is the measure of how much impact the endogenous construct will have if an exogenous construct was removed from the model. Table (4) shows the cut-off for the effect sizes. It can be noticed that course structure has no effect on student satisfaction, while both online tutorials flexibility and technology quality have moderate effect on student satisfaction. Finally, student satisfaction has high effect on student commitment.

Q^2 value indicates the model's out-of-sample predictive power. When a model is said to have a predictive power or predictive relevance, it means that it can accurately predict data not used in the model estimation. The Q^2 value is calculated through running a blindfolding procedure. Before running this procedure, an omission distance (D) must be specified.

Researchers suggest specifying a D between 5 and 10 while being careful that the sample size divided by the selected D would not produce an integer. The omission distance indicates that while running the blindfolding procedure, every x data point of the items will be omitted and then predicted, with x being the specified D value (Hair et al., 2016). Based on the recommendation from the literature, an omission distance of 10 was selected to examine the predictive power of the model. Table (6) presents the Q^2 values obtained from the analysis. The values of Q^2 for

both student satisfaction and student commitment are higher than 0, so it can be safely concluded that the model has a good predictive relevance.

Tenenhaus et al. (2005) proposed the Goodness of Fit (GoF) as a global fit indicator; it is the geometric mean of both the average R^2 and the average variance extracted of the endogenous variables. The GoF index can be calculated as follows:

$$GOF = \sqrt{R^2 \times AVE} = \sqrt{0.5875 \times 0.7118} = 0.647.$$

The criteria of GoF for deciding whether GoF values are not acceptable, small, moderate, or high to be regarded as a globally appropriate PLS model, have been given in table (3). According to these criteria and the value of the GOF (0.647), it can be safely concluded that the GoF model is large enough to consider valid global PLS model.

Conclusion

The present study used structure equation modeling to investigate the behavior of students' commitment towards e-learning during the pandemic of COVID-19 and the key determinants of student satisfaction in online learning environment. Few studies have been developed to examine levels of satisfaction and commitment of students in the Egyptian higher education context through the digitization era.

Over the examination of a sample of 410 students enrolled in an Egyptian higher education institution, the study proved the validity of all the indicators of the variables in the proposed model. Moreover, all variables had a significant impact on student satisfaction except course structure; there was a positive correlation between student satisfaction and student commitment while course structure had an insignificant impact on student commitment. In addition, the majority of students' responses were from College of Management and Technology while other colleges were of same responses and this can justify that even disciplines can differ from one another in the satisfaction and commitment level. Therefore, based on the previous literature review, all variables had significant influence on

both student satisfaction and commitment except the variable named course structure.

Though the present study provides significant information for students and decision makers concerned with levels of student satisfaction and commitment in online learning environment, it was limited to be applied on a specific higher education institution inside Egypt. To address this limitation, the study should be expanded to other higher education institutions inside Egypt in different settings. It will be beneficial and useful for academics and decision makers to understand the behavior of students concerned with the commitment towards their institutions.

Future exploration and investigation might embrace testing the model on other higher education institutions to determine its validity in forecasting the satisfaction and commitment of students in an online learning environment. Some studies (Sun and Zhang, 2006) suggest that moderator variables such as age and gender add a great value in which they can be tested in the future (HASSAN et al.). Furthermore, more variables can be tested to explore their effect on both satisfaction and commitment.

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