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The mediating role of digital informal learning in the relationship between students' digital competence and their academic performance

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ABSTRACT

Students' digital competencies are important for their academic performance. Although scholars have highlighted the importance of students' digital informal learning in developing their digital competence, the mediating role of digital informal learning between digital competence and academic performance have remained unexplored. Thus, this study aimed to investigate the mediating role of digital informal learning between higher education students' digital competence and their academic performance. The data were collected from 319 students from Shiraz University in Iran. Data were analysed using structural equation modeling via AMOS. The result showed a positive effect of the students' digital competence on their digital informal learning and academic performance. Moreover, digital informal learning as the mediator variable had a positive effect on the relationship between digital competence and students' academic performance. In conclusion, to improve students' academic performance, educators and curriculum designers should consider both their digital competence and their digital informal learning.

1. Introduction

The rapid development of digital technologies has spurred strong interest among scholars to employ these technologies for educational purposes, not only to facilitate learning in formal settings (Farrokhnia et al., 2019; Hassanzadeh et al., 2016; Heidari et al., 2020; Noroozi et al, 2016, 2018; Seo et al., 2021; Winkler et al., 2021) but also in informal, out of the class, settings (Chan et al., 2015; Coll & Treagust, 2015; Hubbard, 2019; Song & Lee, 2014). With the recent advent and affordance of new technologies, increasing attention has been paid to digital informal learning (DIL) and the way students learn in such settings (Reinders & Benson, 2017). In DIL, there are no time and space restrictions (Yang, 2020). Thus, DIL provides learners with new opportunities to learn anywhere at any time (Ang et al., 2018, pp. 87–101; Jin et al., 2019; Jurkovič, 2019). DIL is driven primarily by learners' high interests and excitement because they are more motivated to learn in this technology-based learning environment (Sackey et al., 2015) and thus can better improve their academic performance (Doleck et al., 2019; Mishra, 2020).

Academic performance, as one of the most important indicators of success in universities (Mthimunye & Daniels, 2020), refers to

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students' ability to perform classroom tasks, problem-solving activities, and academic affairs (Hung, 2005; Yu et al., 2010). Although scholars reported that academic performance could be positively influenced by students' DIL experiences (Ali & Naylor, 2010), the empirical research on DIL is still insufficient (He & Li, 2019), especially regarding its effect on higher education students' academic performance (Baker, 2014). The picture is also unclear when it comes to influential factors that may regulate the DIL process and outcome (Chan et al., 2015). For instance, scholars claimed that DIL demands proper digital competence (Hubbard, 2019; Nyikes, 2018) to form students' optimal learning and behavior in such a digital-based learning environment (He et al., 2018). Digital competence, as one of the key competencies for lifelong learning (see Ferrari, 2013), is supported by digital literacy (Martin & Grudziecki, 2006). Digital literacy is often referred to as media literacy focusing more on its functional dimension. Digital functional literacy or technical skills means having the practical skills needed to use digital technologies, including operational and creative skills (Falloon, 2020; Helsper et al., 2015). Although digital literacy is often used as a synonym for digital competence (Calvani et al., 2012; He & Zhu, 2017; Ilomäki et al., 2011), they have distinct roots and meanings (Iordache et al., 2017). Digital literacy refers more to a set of skills and activities to achieve digital competence in ICT (Ferrari et al., 2014). According to scholars, the scope of digital competence is beyond digital literacy (Fraillon et al., 2014; Jin et al., 2020) and includes attitude and mindset besides the skills (Janssen et al., 2013). Recent studies suggest digital competence should be replaced with digital literacy in educational contexts since digital competence pays more attention to the ethical, safety, and social dimension (Foulger et al., 2017; Lund et al., 2014) and identifies more diverse knowledge, abilities, and desires of individuals (Falloon, 2020).

Despite the importance of digital competence, so far, only a few studies have investigated the role of students' digital competence in their DIL experiences. For example, Han and Yi (2019), examined the effects of higher education students' digital competence and technology expectancy and how these two interact with cultural differences on DIL. Their results highlighted the essential role of digital competence and technology expectancy in DIL. Moreover, they showed that students' cultural differences play a mediating role in the relationship between their digital competence and DIL behaviors. Pagani et al. (2016), in their study with secondary school students, reported that digital competence had a positive impact on students' academic performance. However, these studies have been conducted separately, and to our knowledge, no study has yet investigated the role of digital competence in both their DIL and academic performance, especially with the focus on higher education students.

As a result, the primary purpose of this study was to investigate the mediating role of DIL between digital competence and academic performance. Based on the reviewed literature (e.g., Baker, 2014; He & Li, 2019; He & Zhu, 2017; Hubbard, 2019; Mangiri et al., 2019), we postulated that there might be an asymmetric relationship between these three constructs. Thus, we first dive into the related literature to formulate different hypotheses related to the key constructs of the study. Afterward, these hypotheses are analysed by structural equation modeling for revealing possible relationships.

2. Research hypotheses and model

2.1. Digital informal learning and academic performance

Learning can occur in a formal or informal context. According to Meyers et al. (2013), formal learning occurs in a structured and organized educational environment and is explicitly designed in terms of time, objectives, and resources. From the learner's perspective, formal learning is intentional, which leads to degrees and certifications (Czerkawski, 2016). In contrast, informal learning occurs relatively in an unstructured and unorganized environment, mostly outside the formal classroom, and it does not have an objective in terms of learning degrees (Czerkawski, 2016; Meyers et al., 2013). In informal learning settings, the control of learning is in the learner's hands (Callanan et al., 2011; Falk & Dierking, 2002), allowing them to have more control over the opportunity to learn, selecting what to learn, and evaluating themselves than formal learning setting (He & Li, 2019; Huang & Oh, 2016; Laurillard, 2009). The use of these opportunities has increased with the entrance of technology into life, especially in educational fields, and has provided a platform for a new type of learning called DIL (Dron & Anderson, 2014; Holland, 2019; Lai, 2017).

DIL has been defined as the unstructured technology-mediated learning opportunities in informal learning settings (He et al., 2018; Ungerer, 2016). One of the most common settings for DIL is the Internet that provides further opportunities for searching information and finding the answers to learners' problems (Lai, 2017). Similar to any learning behavior, DIL has different aspects, namely cognitive learning (CL), metacognitive learning (MCL), and social and motivational learning (SML) (He & Li, 2019). Based on the cognitive aspect, learners manipulate digital media to learn mentally or physically (Mayer, 1998). MCL refers to the process of planning, organizing, storing knowledge, monitoring comprehension, and facilitating learning to perform cognitive tasks (Mayer, 1998; Vermunt, 1996). MCL influences various learning strategies that learners choose, such as recognizing what approaches are better for memorizing information (Taheri et al., 2020). In the case of DIL, learners should choose the most appropriate strategy or technology that provides a better learning opportunity for them (Vermunt, 1996). Finally, the social and motivational aspects respectively refer to the essential social interaction in the digital setting to collaboratively build new knowledge or getting motivated to learn (Vermunt, 1996; Mayer, 1998).

Previous research confirmed the positive effect of DIL on students' learning outcome (e.g., Lee & Dressman, 2018), engagement (e. g., Pechenkina et al., 2017), and retention of domain-specific knowledge (e.g., Pechenkina et al., 2017; Sommerauer & Müller, 2014) in different educational contexts. One of the students' learning outcomes is academic performance. Some previous studies reported a positive effect of informal learning experiences on students' academic performance (e.g., Baker, 2014; Cox, 2013; Sackey et al., 2015). However, other studies reported negative findings related to the effect of students' DIL experiences on their academic performance. For instance, some scholars showed that using social media such as Facebook for learning purposes had a negative effect on students' GPA (Junco, 2012b; Junco & Cotten, 2012; Kirschner & Karpinski, 2010) and negatively predicted their engagement scale scores (Junco,

2012a). In the same vein, Alloway et al. (2013) findings indicated that young people who used Facebook for more than a year had higher scores in tests of verbal ability, working memory, and spelling compared to their peers who had used it for a shorter period; but there was no difference between these two groups for math and cognitive scores.

Therefore, by considering these contradictory findings and since, to date, no study has investigated the direct effect of DIL on higher education students' academic performance, the following hypothesis was formulated for this study:

H1. DIL has a positive effect on higher education students' academic performance.

2.2. Digital competence, digital informal learning, and academic performance

According to Ferrari (2012), people with digital competence can use digital media to search for information, and they can critique what they receive from cyberspace, and at the same time gain the ability to communicate with others using a variety of digital tools and applications such as mobile phones or social media. Thus, digital competence is a multi-dimensional ability that can include ICT knowledge, ethical awareness, and higher-order cognitive skills (Calvani et al., 2012). Scholars refer to various theoretical frameworks for digital competence in the literature (Van Laar et al., 2017). For instance, based on Carretero et al. (2017), the digital competence framework has gone through three periods (i.e., DigComp 1.0, 2.0, and 2.1). DigComp 2.1 includes 21 sub-competencies structured to improve citizens' digital competence, consisting of five dimensions: information and data literacy, communication and collaboration, digital content creation, digital safety, and problem-solving. Each of the 21 sub-competencies is accompanied by descriptors of eight proficiency levels as well as examples of the requisite knowledge, skills, and attitudes for these eight levels. This framework is designed for developing the citizens' digital competence (Carretero et al., 2017).

Calvani et al. (2009, 2012) also introduced a digital competence framework that includes three technical, cognitive, and ethical dimensions. The technical dimension means that the user can discover and deal with new problems and technology context in a flexible way. This dimension consisted of visual literacy, troubleshooting, and understanding technological concepts. The cognitive dimension means that the user can read, select, interpret, and evaluate data and information, considering the information's appropriateness and reliability. This dimension includes organizing and connecting textual and visual data, organizing structured data, and information research. The ethical dimension means the ability to interact constructively as well as the responsibility with other people in using technologies that consist of staying safe online and respect for others (Vieru et al., 2015). Similar to the DigComp framework, Calvani et al.'s (2012) digital competence framework also used Bloom's taxonomy of learning and described the knowledge, skills, and attitudes underlying digital competence in five areas of competence (Ferrari et al., 2014).

This study adopted Calvani et al.'s (2012) digital competence framework for two reasons: Firstly, this framework is multidimensional, which shows that the concept of digital competence is non-linear and is a combination of abilities, cognitive, relational, and social skills. Secondly, this framework is interconnected, which means that it is dependent on other abilities or basic competencies (e.g., reading, problem-solving, logical, inferential, and metacognitive skills) that overlap with digital competence (Calvani et al., 2012).

Some scholars put a relatively high emphasis on the positive effect of students' digital competence on their learning and performance in digital formal learning environments (e.g., Elstad & Christophersen, 2017; Martín & Tyner, 2012; Nyikes, 2018). Scholars also claimed that digital competence could have a substantial role in optimizing DIL's positive effects on students' learning outcomes (e.g., Hatlevik, Guðmundsdóttir, & Loi, 2015; He et al., 2018; Hubbard, 2019; Pogue, 2019). Besides, a considerable body of literature highlighted the importance of digital competence in learners' performance at different educational levels, such as for teachers (Mangiri et al., 2019), university faculty members (Yazon et al., 2019), and middle school students (Hatlevik, Ottestad, & Throndsen, 2015).

In addition to the above-mentioned studies, there are various findings about the effect of ICT on academic performance. Some studies reported a positive relationship between the use of ICT and academic performance (e.g., Basri et al., 2018; Nketiah-Amponsah et al., 2017; Çakıroğlu et al., 2017). Other studies showed that moderate access and use of ICT could increase academic performance, but over-access to ICT resources, use, and interest in ICT was associated with declining academic performance (e.g., Gubbels et al., 2020; Park et al., 2014). Fraillon et al. (2019) reported that students' daily use of computers at home and experience with computers was consistently related to computer and information literacy. However, they found no clear relation between the use of ICT for school-related purposes and the computer and information literacy score. In another study, Fraillon et al. (2014) declared that higher ICT development index scores were typically associated with higher computer and information literacy scores across countries. However, in contrast to these findings, Peña-López (2015) reported that the most frequent pattern that emerged in the Programme for International Student Assessment's (PISA) data showed a weak or sometimes negative association between investment in ICT use and students' academic performance.

However, in general, empirical studies that have investigated the effect of higher education students' digital competence on their DIL are scarce, especially with regard to their academic performance (He et al., 2018; Meyers et al., 2013). Only a few studies investigated the *in-direct* effect of digital competence together with students' technology expectancy (He & Li, 2019) and innovativeness and attitudes (He & Zhu, 2017) on DIL outcomes and adoption. Therefore, by considering these research gaps and previous literature, the following hypotheses are formulated to be investigated in this study:

H2. Higher education students' digital competence has a positive effect on their DIL.

H3. Higher education students' digital competence has a positive effect on their academic performance.

2.3. The mediating role of digital informal learning

The results of Kim et al.'s (2018) study confirmed the positive effect of learners' prior experiences of using technologies on acquiring digital competence and their willingness to use technologies. Similarly, Khalifeh et al. (2020) reported that higher education students with a personal computer, laptop, or tablet demonstrated higher levels of readiness for online learning environments in their study on higher education students. Therefore, this study puts forward the idea that DIL can moderate the probable relationship between students' digital competence and academic performance. In the sense that students' DIL can contribute to their academic performance in two ways, first by a direct effect on their new knowledge acquisition outside the formal classes, and second by improving their digital competence, which may further contribute to their academic performance and success. As a result, the following hypothesis can be formulated and will be investigated in the current study:

H4. DIL plays a mediating effect in the relationship between higher education students' digital competence and their academic performance.

The path diagram related to the hypothesis of this study is shown in Fig. 1.

3. Method

3.1. Study design and sample

This study was explanatory by nature. The study was conducted at Shiraz University, one of the largest comprehensive higher education institutes in Iran, in the academic year 2019–2020. Based on the Cochran formula for a finite population (see Cochran, 1977), 350 students were recruited to participate in this study with 95% confidence and a 5% error level, which was also determined as an adequate sample size based on Westland's (2010) formulas. Moreover, with the effect size of 0.18, the desired statistical power level of 0.8, the probability level of 0.05, the number of latent variables of 3, and the number of observed variables of 10, the minimum sample sizes for the AMOS structural equation modeling analysis was calculated to be 156, which was much smaller than our sample size of 350 respondents. The recruited students were required to answer the questionnaires through a stratified sampling method that the marker variables were level of degree and fields of study. However, only 319 questionnaires were returned, which indicated that the response rate was 91%. As nonresponding participants were randomly distributed in different levels of degree and fields of study, the obtained data set was not biased. Table 1 shows the demographic information of the samples.

3.2. Data analyses

The SPSS data analysis software (version 22) was used to examine the validity of the instruments and also to provide descriptive analysis of the data. For structural equation modeling and also for model fitness, the AMOS module (version 21.0) was used. According to Lei and Wu (2007), one of AMOS's capabilities is producing bootstrapped standard error estimates, together with the maximum likelihood estimation method. The maximum likelihood method to estimate the model's parameters was also adopted in the current study, where all analyses were conducted on variance-covariance matrices (see Hair et al., 1998). Besides, to evaluate the overall fit of the model, several recommended indices were calculated in the present study: chi-square statistic (χ 2), χ 2/df ratio, the incremental fit index (IFI), the comparative fit index (CFI) (Hooper et al., 2008), and the root-mean- square error of approximation (RMSEA) (Browne & Cudeck, 1989; Hooper et al., 2008). The acceptable threshold level of the model fit indicators is different in the various resources. For example χ 2/df ratio <3, IFI and CFI >0.90, RMSEA <0.08, SRMR <0.10 (Browne & Cudeck, 1989; Hair, 2010); SRMR < 0.08 and CFI > 0.95 and SRMR < 0.08 (Hooper et al., 2008; Kline, 2015); RMSEA < 0.07 (Hooper et al., 2008) and RMSEA < 0.05 (Kline, 2015). As a result, CFI between 0.90 and 0.95, RMSEA between 0.05 and 0.08, SRMR between 0.08 and 0.10, and RMSEA between 0.05 and 0.08, and CFI between 0.90 and 0.95 indicates the adequate level for the current study. For mediation analyses, we used bootstrapping with 2000 bootstrap resamples to examine indirect effects in mediation models. Confidence intervals that do not contain zero indicate a significant indirect effect (mediation).

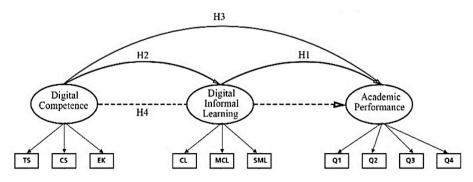


Fig. 1. Path diagram related to the study hypothesis.

Table 1

Characteristics of the sample.

Items	Categories	Number	Percentage
Gender	Male	181	56.7
	Female	138	43.3
Level of degree	Undergraduate	186	58.3
-	Master	89	27.9
	Ph.D	44	13.8
Fields of study	Social sciences and humanities	162	50.8
	engineering	94	29.5
	Basic sciences	63	19.7

3.3. Measures

3.3.1. Digital competence

The instrument designed by He and Zhu (2017) was used to measure students' digital competence. He and Zhu (2017) developed this questionnaire based on Calvani et al.'s (2012) digital competence framework. This instrument includes three dimensions, i.e., technical skills (TS), cognitive skills (CS), and ethical knowledge (EK). The TS dimension includes 11 items (q1 to q11) related to the basic technological skills of students, such as "I can use some software programs to deal with data visualization." The CS dimension also includes 12 items (q12 to q23) regarding the needed cognitive skills for using technology, such as "I have a strong awareness of reliability of information when searching online." Finally, the EK dimension consists of 6 items (q24 to q29) related to the ethical issues of using technology, for instance, "When I deal with online payment, I am always aware of the safety issue." Each item includes 5-point Likert scale (1 = strongly disagree; 5 = strongly agree).

Questionnaire designers with different colleagues in the various studies have examined and confirmed the questionnaire's validity, especially by convergent and discriminant validity (He et al., 2018, 2020; He & Li, 2019; He & Zhu, 2017). In this study, the instrument validity was also examined by calculating the confirmatory factor analysis (CFA). Based on Buyukozturk (2007), the substance's factor load must be at 0.30 and above. Therefore, six items, i.e., q10, q18, q24, q26, and q27, were removed from the scale since their factor loads were below 0.30. After eliminating the items, the analysis scale and its dimensions were rebuilt. Finally, factor loadings of TS items were between 0.31 and 0.73; CS items were between 0.35 and 0.75, and EK items were between 0.47 and 0.80. Therefore, each dimension was formed based on its remaining items' average scores after removing the low score items. Furthermore, the fit index of the scale was as follows: $\chi 2 = 577.02$, df = 254, P-value = 0.0001, $\chi 2/df = 2.27$, IFI = 0.90, CFI = 0.90, RMSEA = 0.06 and SRMR = 0.01.

In addition, the reliability of the instrument was estimated according to the CFA. The overall reliability of the instrument was calculated, and Cronbach's alpha value was 0.90. The calculated Cronbach's alpha of each dimension, i.e., TS (0.84), CS (0.83), and EK (0.80), confirmed the high reliability of this instrument for measuring students' digital competence. Also, the composite reliability (CR) for each dimension was obtained: TS (0.83), CS (0.83), and EK (0.73). These high composite reliability scores were above the critical value recommended by scholars (e.g., Hair et al., 1998).

3.3.2. Digital informal learning

He and Li's (2019) questionnaire was used to measure students' DIL. This questionnaire included three dimensions, i.e., CL, SML, and MCL. The CL dimension included four items related to the cognitive goals of using digital technologies in informal learning, such as "I often use digital technologies to engage in self-expression." The CL dimension was formed based on the average scores of q1 to q4. The SML dimension included four items related to the social and motivational goals of using technologies in informal learning, such as "I often use digital technologies to engage in constructive activities." and was formed based on the average scores of q9 to q12. Finally, the MCL dimension consisted of 4 items that measure DIL's meta-cognitive aspects, such as "I often use digital technologies to expand learning opportunities." The MCL dimension was formed based on the average scores of q8.

The questionnaire's validity have been confirmed in various studies by questionnaire designers and their colleagues (He et al., 2018, 2020; He & Li, 2019; He & Zhu, 2017). In addition, the reliability of this tool has also been investigated in different cultures. For example, in China, the Cronbach's alpha for CL, SML, and MCL, were 0.89, 0.89, and 0.90, and in Belgium, they were 0.79, 0.80, and 0.83 respectively (see He & Li, 2019). In the current study, the validity of the instrument was examined by calculating CFA. Factor loadings of CL items were between 0.46 and 0.79, SML items were between 0.48 and 0.77, and MCL items were between 0.64 and 0.83. The fit index of the scale was as follows: $\chi 2 = 129.60$, df = 49, P-value = 0.0001, $\chi 2/df = 2.64$, IFI = 0.95, CFI = 0.95, RMSEA = 0.07 and SRMR = 0.04. The reliability of the instrument was examined according to the CFA by calculating the Cronbach's alpha. Cronbach's alpha coefficients for DIL was 0.89, and for its diminutions were CL (0.75), SML (0.82), and MCL (0.78). Moreover, the values of CR for each dimension respectively were obtained: CL (0.72), SML (0.82), and MCL (0.74), which was above 0.70.

3.3.3. Academic performance

Students' academic performance has been measured differently in literature. Some studies measured academic performance by students' grade point average or GPA (e.g., Kirschner & Karpinski, 2010; Lau, 2017). Others measured academic performance by students' self-reported data such as students' perception of their ability to learn, perform academic tasks and their overall academic performance (e.g., Chang et al., 2019; El Ansari et al., 2020; Escalante Mateos et al., 2021; Han & Yi, 2019; Martin et al., 2005;

Richardson et al., 2005). In this study, academic performance was also measured by students' self-perception using Yu et al.'s (2010) questionnaire. This instrument included four items of students' academic performance, such as "I have learned how to perform my coursework efficiently successfully." The validity of this instrument has also been examined and confirmed in various studies (Nkhoma et al., 2015; Yu et al., 2010; Zhou, 2017). The validity of the instrument was also examined in the current study by calculating CFA. Factor loadings of items were between 0.50 and 0.86. The fit index of the scale was as follows: $\chi 2 = 6.5$, df = 2, P-value = 0.03, $\chi 2/df = 3.25$, IFI = 0.90, CFI = 0.90, RMSEA = 0.08 and SRMR = 0.02. According to the CFA, the reliability of this instrument was confirmed in the present study (Cronbach's $\alpha = 0.86$). Also, the value of CR was obtained 0.82, which was above 0.70.

4. Results

4.1. Preliminary analyses

Table 2 demonstrates the means, standard deviations, skewness, kurtosis, Cronbach's alpha, and correlations between all the research variables. The means scores of digital competence dimensions ranged from 3.39 to 3.45, the means scores of DIL dimensions ranged from 3.68 to 3.76, and the mean score of academic performance was 3.44, which were higher than the average level (3) based on a 5-point Likert scale, and they were in the third quartile (Q3). The standard deviations ranged from 0.68 to 0.82. Skewness and kurtosis were used to check the normality of the constructs. All variables satisfied skewness (below 2) and kurtosis (below 7) as a multivariate normal distribution (Hu & Bentler, 1999). These findings revealed that the model was suitable for SEM analysis. There was a positive correlation between the dimensions of the DIL variables, the dimensions of the digital competence variables, and academic performance.

4.2. Measurement model

We evaluated construct reliability and convergent validity for the measurement model. Appendix 1 provides information on the reliability and convergent validity, showing that the first- and second-order measurement models are correct. Construct reliability was assessed with CR. Cronbach's alpha and CR values were above the threshold value of 0.70 (Fornell & Larcker, 1981). For the first-order measurement model, the CR values ranged from 0.75 to 0.89, which were above 0.70. The Cronbach's alphas of all variables were higher than 0.86, which were above the suggested value of 0.70. The convergent validity of the measurement model was also calculated by confirmatory factor analyses. The results showed that the factor loadings dimensions of the latent variables (digital competence, DIL, and academic performance) on the measurement model ranged from 0.51 to 0.89. The fit index of this model was sufficient, as well ($\chi = 67.80$, df = 32 P-value = 0.0001, $\chi 2/df = 2.11$, IFI = 0.97, CFI = 0.97, RMSEA = 0.05 and SRMR = 0.05)."

In the second-order, latent variables such as TS, CS, EK, CL, MCL, and SML were specified with all of their underlying items. For these variables, the CR values ranged from 0.72 to 0.83. Cronbach's alpha of all constructs was above 0.75. The convergent validity of the second-order latent variable was calculated by confirmatory factor analyses. The result showed that the factor loadings of the latent variables were significant as follows: For the digital competence variable, the TS dimension 0.33–0.75, the CS dimension 0.43–0.73, and the EK dimension 0.61–0.75, for the DIL variable, the CL dimension 0.48–0.78, the MCL dimension 0.62–0.84, and the ML dimension 0.48–0.76, and finally for the academic performance items were between 0.52 and 0.83. Based on Buyukozturk (2007), the factor load of the substance needs to be at 0.30 and above. Therefore, *five* items of TS and EK dimensions (q10, q18, q24, q26, and q27) were removed (see Appendix 1). The fit index of this model was also good ($\chi 2 = 1267.37$, df = 696 P-value = 0.0001, $\chi 2/df = 1.82$, IFI = 0.90, CFI = 0.90, RMSEA = 0.05 and SRMR = 0.07).

4.3. Structural model

We analysed the structural model to investigate the research hypotheses. As hypothesized, the structural model offered a good fit to the data ($\chi 2 = 46.130$, p = 0.0001; $\chi 2/df = 1.53$; IFI = 0.98; CFI = 0.98; RMSEA = 0.04; SRMR = 0.036). The findings indicated that increasing DIL had a positive effect on improving students' academic performance ($\beta = 24$, p = 0.0001), which confirmed our first research hypothesis (i.e., H1). The findings also showed that increasing students' digital competence had a positive and significant effect on increasing DIL ($\beta = 0.59$, p = 0.0001), which confirms the second hypothesis (i.e., H2). Moreover, the results of the analysis showed that increasing students' digital competence had a positive effect on improving their academic performance ($\beta = 0.46$, p = 0.0001). So, the third hypothesis of the study, i.e., H3, was also confirmed. Table 3 demonstrates the summary of the direct, indirect,

Table 2	
Means, standard deviations, correlations,	and reliabilities.

1 2 3 4 5 6 7 (0.84)
0 (7** (0.00)
0.67** (0.83)
0.37** 0.43** (0.80)
0.23** 0.34** 0.29** (0.75)
0.27** 0.34** 0.30** 0.64** (0.82)
0.23** 0.31** 0.25** 0.57** 0.66** (0.78)
0.28** 0.39** 0.37** 0.45** 0.33** 0.33** (0.80
0 0 0 0

Table 3

H#	Proposed Relationship	Direct Effect	Indirect Effect	Total Effect
H1	DIL	0.24	-	0.24
	Academic performance			
H2	Digital Competence	0.59	-	0.59
	DIL			
H3	Digital Competence	0.46	-	0.46
	Academic performance			
H4	Digital Competence	0.46	0.14	0.60
	DIL Academic			
	> Performance			

and total effect of proposed relationships.

4.4. Mediation analyses

The Bootstrap method with 2000 resamples was used to investigate the mediating effect and the indirect effect of DIL in the relationship between students' digital competence and their academic performance. The bootstrap method was used since it has been advocated in recent methodological publications as the preferred approach to mediation analysis, especially in the case of large samples (see Hayes et al., 2011; Rungtusanatham et al., 2014; Zhao et al., 2010). The results showed that the direct effect of digital competence on increasing academic performance was $\beta = 0.46$, p = 0.0001. The indirect effect of digital competence on increasing students' academic performance with mediating role of DIL was $\beta = 0.14$, wherein the CI was between the range of 0.01–0.23, and the 95% CI did not include zero. This result indicates that the indirect effect between digital competence and academic performance through DIL is significant.

After including DIL as an intermediate variable, the association between digital competence and academic performance was reduced, but it still remained significant. This shows that DIL partially mediated the effect of digital competence on academic performance; hence hypothesis H4 was supported. The indirect effect of digital competence on academic performance through DIL was significant. Thus, considering the confirmation of the research hypotheses and the significant effects of the research variables on each other, it can be concluded that the proposed research model was approved as the final model (see Fig. 2).

5. Discussions

This study aimed to identify the mediating role of DIL between higher education students' digital competence and academic performance. In general, the study results indicated that DIL has a positive effect on students' academic performance. The analysis also confirmed that digital competence has a significant positive effect on students' DIL and academic performance. Moreover, DIL experiences contributed to students' academic performance in two ways, (1) by a direct effect on their performance and (2) by improving their digital competence.

According to the first hypothesis, the findings showed that the DIL has a positive effect on higher education students' academic performance. Thus, we can argue that DIL environments can better enable students to seek learning strategies, monitoring their learning progress, expanding their learning opportunities, and acquiring comprehensive knowledge. Therefore, these environments can genuinely improve students' academic performance besides their academic classes. In the same line, Petrovic (2019) pointed out that an important part of human life learning takes place outside the educational system, unintentionally, and through digital

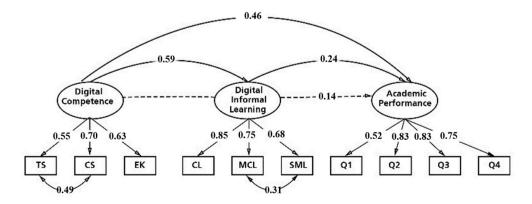


Fig. 2. Path analysis depicting direct and indirect effects of variables on each other.

technologies. The finding of this study is also in line with previous studies for different learning outcomes and educational contexts (Trinder, 2017). For example, Huang and Oh (2016) reported that Internet-based informal learning could affect students' learning and academic performance and facilitate their decision-making processes. Also, Chang et al. (2019) showed in their studies, students with higher academic performance are more likely to use the Internet to relax or reduce their study load with less science but entertainment. Furthermore, they argued that DIL sustains students' motivation in learning, engages them in constructive activities, and teaches them how to perform their coursework successfully. Lee and Dressman (2018) showed that participation in a variety of DIL of English activities significantly predicts students' English language proficiency. Their result also showed that participants' diverse use of DIL of English contributed to a higher rating of willingness and motivation toward learning English. Align with their finding, He and Li (2019) argue that Deci and Ryan's (1985) self-determination theory can serve as a theoretical basis for DIL. According to self-determination theory, there is a positive link between the facilitation of a certain degree of autonomy and the degree of engagement and motivation that learners feel (Proulx et al., 2017). The autonomy that DIL provides for the students (see He & Zhu, 2017) will ultimately increase their engagement with the learning goals and process, which this engagement will then contribute to their academic performance (Lee, 2014). Therefore, not only formal learning but also DIL can improve students' academic performance in one way or another.

Regarding our second hypothesis, the findings showed a positive and significant effect of digital competence on students' DIL, in the sense that the higher the competencies of students, the better they will be able to learn and achieve their goals in DIL. This result is consistent with the previous studies (Elstad & Christophersen, 2017; Martín & Tyner, 2012; Nyikes, 2018). Jin et al. (2019), in their review study related to the adults' informal learning using mobile devices, concluded that adults learn more from informal learning supported with mobile technology since it enables them to access content and make more interactions at any time and place according to their preferences. Also, considering the mediating effect of DIL, the more students use DIL opportunities, the higher their digital competencies, which further contribute to their academic performance. Students' positive prior digital experience significantly influences their digital competence and their attitude toward using digital technologies for educational purposes (Kim et al., 2018).

Similar to the previous hypotheses, the result of this study confirmed our third hypothesis, indicating that when students' digital competence is improved, their academic performance will also be improved. Many studies confirm the positive relationship between digital competence and academic performance in different educational contexts (Hatlevik, Ottestad, & Throndsen, 2015; Kim et al., 2018; Mangiri et al., 2019; Yazon et al., 2019). For instance, in vocational high schools, Mangiri et al. (2019) reported that teachers' digital competence plays a decisive role in their professional development. Also, in higher education, students should improve their computer literacy skills in line with the rapid global digitalization. Acquiring digital competence allows students to obtain technical, cognitive, and ethical skills and knowledge. Also, digital competence helps students design their plan or schedule related to their scientific needs, solve software problems by searching online, communicate rationally with others online, and find the answer to their scientific questions and problems with the help of their digital competence.

6. Conclusion, limitations, and implications

The results of this study showed positive mediating role of DIL in the relationship between students' digital competence and their academic performance. Although digital competence has a direct positive effect on students' academic performance, together with DIL experiences, it yields a greater effect on their performance in their formal classes. Since the DIL plays a mediating role in the relationship between students' digital competence and their academic performance, digital competence can directly affect academic performance, as well.

Despite the important result of this study for higher education, several limitations of this study should be acknowledged. In this study, similar to other studies (Aesaert et al., 2017; Chang et al., 2019; El Ansari et al., 2020; Escalante Mateos et al., 2021; Han & Yi, 2019; Martin et al., 2005; Richardson et al., 2005), we used self-reported data for measuring students' digital competence and their academic performance. Aesaert et al. (2017) shows that students make accurate judgments about their ability in digital information processing and communication, meaning that the magnitude of students' judgment error of their competencies is not big. However, students in self-reported measurement instruments may have a bias regarding their experiences. For example, they may provide responses that are socially acceptable in such self-reported instruments (Devaux & Sassi, 2016). Next to social desirability, in self-reported measurements, over-confident students may overestimate their competencies as opposed to less-confident students who may underestimate their competencies (see Aesaert et al., 2017; Noroozi et al., 2013). Therefore, in order to deal with the flaws that are inherent to self-reported measurements, we suggest future studies to use both self-reported and actual performance tests simultaneously to find out the extent to which the findings of self-reported data (perceived academic performance) are in line with the actual performance (real academic performance) of students.

Another important limitation is the methodology of the current study. This study focused on the complex relationships between various research variables using structural equation modeling without any empirical intervention. Therefore, future research can focus on specific strategies, for example, for developing students' digital competence and DIL to investigate their effects on students' academic performance in empirical settings. Also, some research can explore the strategies that can make students use DIL.

Based on the findings of this study, in this digitalization age, educators and educational policy-makers should pay more attention to promoting and improving students' digital competencies in order to enhance their academic performance. Also, as students have more flexibility and control over their learning in DIL settings, university policy-makers should pay attention to informal learning contexts alongside formal learning. Also, in order to improve students' academic performance, higher education institutions, educators and curriculum designers should not only improve the students' digital competence through extra-curricular programs (Basantes-Andrade et al., 2020), but also should provide opportunities for students to learn outside the formal classes through digital environments.

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Appendix 1. Confirmatory factor analysis

factor	loadings	Cronbach's alpha	CR
Digital competence		0.90	0.76
TS 0.75		0.84	0.83
I can use some software programs to deal with data visualization	0.33		
I can use a certain software program to edit pictures	0.57		
I can use a certain software program to make video	0.57		
I can use a variety of programs to deal with Antivirus problem for computer	0.68		
I can deal with computer system problems	0.72		
I can deal with software problems by searching online	0.75		
I can use at least one operating system skillfully (e.g., Windows, OSX, Android, IOS, etc.)	0.70		
I am able to operate at least a personal computer-related hardware (desktop, laptop, tablet device or smartphone device)	0.57		
I am able to use social media well	0.43		
I am able to use some photo and video sharing tool well	_		
I am able to use a variety of APPs	0.42		
CS 0.89		0.83	0.83
I can represent a text with a graph by digital tools	0.71	0.00	0.00
I can represent hierarchical classes with digital tools	0.69		
I can identify the keywords well in a digital text very well	0.70		
I can organize data in a table by a variety of digital tools	0.73		
I can find missing values in a table of digital context	0.73		
с с с с с с с с с с с с с с с с с с с	0.43		
I can use digital technologies to design my plan or schedule			
I am quite confident at searching for information online what I need	-		
I am quite confident at searching for information on a certain database	0.43		
I have a strong awareness of the credibility of information when search online	0.47		
I have a strong awareness of the reliability of information when search online	0.56		
I often consider the reliability of information online when I quote or share it	0.43		
I often consider the credibility of information online when I quote or share it	0.57		
EK	0.51	0.80	0.73
When I surf online, I have awareness of Privacy (personal information, personal Photos, etc.)	-		
When I deal with online payment, I am always aware of the safety issue	0.75		
When I see online content that makes me feel uncomfortable, unsafe or worried, I will leave it or take an action	-		
I often post pictures with my friends on the Internet without their permission (reverse)	-		
I have an awareness of online bullying and behave myself decent online	0.65		
I have the awareness to comment or communicate with others online in a rational way	0.61		
Digital Informal Learning		0.89	0.83
CL 0.77		0.75	0.72
I often use digital technologies to enhance my understanding of school course materials in informal learning contexts	0.62		
I often use digital technologies to expand knowledge of the discipline	0.72		
I often use digital technologies to keep informed of the development in the discipline	0.78		
I often use digital technologies to engage in self-expression	0.48		
		0.82	0.74
I often use digital technologies to seek learning strategies and tips	0.62	0.02	0.74
	0.62		
I often use digital technologies to help myself to monitor learning progress			
I often use digital technologies to expand learning opportunities I often use digital technologies to seek engaging learning experiences	0.84 0.67		
		0.07	0.00
SML	0.77 0.76	0.87	0.82
I often use digital technologies to sustain motivation in learning	0.70		

(continued on next page)

(continued)

factor	loadings	Cronbach's alpha	CR
I often use digital technologies to elicit support and help	0.73		
I often use digital technologies to engage in constructive activities	0.62		
I often use digital technologies to collaborate with people for learning	0.48		
Academic Performance		0.86	0.82
I am confident about the adequacy of my academic skills and abilities	0.52		
I feel competent conducting my course assignments	0.83		
I have learned how to perform my coursework in an efficient manner successfully	0.83		
I have performed academically as well as I anticipated I would	0.75		

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