

## Collaborative learning in online breakout rooms: Effects of learner attributes on purposeful interpersonal interaction and perceived learning

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# Collaborative learning in online breakout rooms: Effects of learner attributes on purposeful interpersonal interaction and perceived learning

#### **Abstract**

**Purpose** – Breakout rooms are commonly used by lecturers as a means to achieve collaborative learning in online lessons. Although breakout rooms can be effective at encouraging student engagement, interaction, and learning, many students dislike being forced to interact with peers, and for some students it can lead to feelings of anxiety and stress. Successful collaborative learning depends upon having the 'right' individuals working together, so the purpose of this research is to identify specific learner attributes that are associated with purposeful interpersonal interaction in breakout rooms.

**Design/methodology/approach** – An online survey was used to obtain data from 664 higher education students in the United States, which were analyzed using partial least squares structural equation modeling (PLS-SEM).

**Findings** – Students' technology readiness, social identification, and intercultural communication competence are each significantly related to the achievement of purposeful interpersonal interaction, which is strongly related to students' perceived learning.

Originality/value – The breakout room represents a unique and specific context for collaborative learning, where there may be minimal lecturer supervision, and where students may choose to disengage by turning off their cameras and microphones, or simply listen without participating (known as lurking). The existing literature has given little attention to how lecturers allocate students to online breakout rooms.

**Practical implications** – The findings of this research emphasize the importance of lecturers considering learner attributes when forming breakout room groups.

**Keywords** Online learning, Collaborative learning, Breakout rooms, Teaching/learning strategies, Group formation

Paper type Research paper

#### Introduction

The COVID-19 pandemic that started in early 2020 forced universities globally to make more use of online learning. Institution managers determined that face-to-face classes be suspended, and lecturers were instructed to consider how best to deliver online lessons to ensure that students' learning continued with minimal disruption. As many countries transition into a post-pandemic era, the flexibility and opportunities provided by online technologies are likely to remain part of students' learning experiences in higher education. Some students struggle with online teaching and learning because they do not have reliable internet access (Bento, 2022). Other students may not be mentally technology ready, or they may lack confidence or the necessary interpersonal and communication skills to effectively engage with peers and lecturers in online settings (Tang *et al.*, 2021). Chaw and Tang (2022) found that learner characteristics may affect individuals' preferences for particular types

of learning environment. This research assesses the extent to which specific learner attributes may influence individuals' attitudes and readiness for online collaborative learning. Effective collaborative learning requires high quality, organic, and valid communication exchanges between students, which directly relate to the achievement of established learning outcomes or to the building of social relationships. Such exchanges may be regarded as purposeful interpersonal interaction.

Lecturers need to structure and deliver lessons to achieve purposeful interpersonal interaction among students (Nikou and Maslov, 2022). During the switch to 100% online teaching, many lecturers were advised by managers, peers, and professional development trainers to use breakout rooms as a means to achieving purposeful interpersonal interaction among students (McGrath and Wolstencroft, 2021). A breakout room is a virtual space that is separate from the main online classroom, in which only the students allocated to the room may hear and participate in discussion, as well as read and write text chat messages to each other (Chandler, 2016). The main objective of a breakout room is to promote collaborative learning, whereby students develop their knowledge and skills through interaction with their peers (Lyons *et al.*, 2021).

Collaborative learning is widely used in both physical and online classrooms. In online settings, the lecturer decides how many students should be in each breakout room, and whether or not to select specific students for each room. Collaborative learning may deliver both individual and group-level benefits (Curşeu and Pluut, 2013). It may encourage student motivation, involvement, and engagement, and it may contribute to achieving deeper understanding of program content and concepts; enhanced overall learning; communication and interpersonal skills development; and higher satisfaction with lesson delivery (Chen and Kuo, 2019; Lyons *et al.*, 2021). Quieter and less confident students are often more willing to participate in breakout room discussions, as these may be perceived as less threatening or intimidating than speaking to the whole class (Chandler, 2016).

Although breakout rooms can be effective at encouraging student engagement, interaction, and learning, they may allow some students to lurk, where they become free-riders, who listen and observe, but not participate (Bozkurt *et al.*, 2020). Collaborative learning generally takes more time than lecturer-led delivery, and it may be less effective when students lack the necessary knowledge and skills to be self-reliant, or when the group has underachievers, disruptive members, or individuals who refuse to interact or engage with others (Smith *et al.*, 2011). Many students dislike being forced to interact with peers, and for some students it can lead to feelings of anxiety and stress (McGrath and Wolstencroft, 2021). Indeed, many students turn off their cameras and microphones while in breakout rooms, which may lead to awkward silences and no purposeful interpersonal interaction that delivers collaborative learning.

Successful collaborative learning in online breakout rooms depends upon having the 'right' individuals working together, who interact with one another in a productive and supportive manner. However, there is little consensus in the literature about how lecturers should create breakout room groups (Chen and Kuo, 2019). Lecturers may allow the students to form their own groups, which are then usually based on friendships, or the desire to work with high ability individuals who will contribute most to achieving the set tasks and learning goals (Marder *et al.*, 2021). Alternatively, lecturers may save time and effort by allowing the online learning platform to randomly assign individuals to groups. Finally, lecturers may determine that it is best to take responsibility for group

formation themselves. When this option is selected, a key decision is whether to create groups comprised of homogeneous or heterogeneous individuals. Regardless of this decision, the lecturer needs to consider relevant learner attributes to make his/her final allocations. Sometimes, lecturers may be able to allocate students based on existing or previous groups, perhaps stemming from earlier courses or projects. Lecturer behaviors are critical in the online classroom as these behaviors may motivate or demotivate students, and adopting a person-centered approach is more likely to result in students being engaged (Gupta, 2022).

The existing literature on student group formation has not specifically considered group formation for online breakout rooms, which represent a unique learning environment that requires students to interact with one another, usually with minimal lecturer supervision. The need to consider learner attributes for breakout room groups provides the rationale for this research, and key learner attributes that may be associated with purposeful interpersonal interaction are examined.

The study has two research questions:

- RQ1 What learner attributes are associated with purposeful interpersonal interaction in online breakout rooms?
- RQ2 Is purposeful interpersonal interaction associated with students' perceived learning?

Through answering these questions, we fill a literature gap that connects the group formation process with the use of online breakout rooms, and provide much needed information about the types of student who are more likely to willingly engage in purposeful student-student interaction. The findings may be useful to lecturers who want to form online breakout room groups that will maximize effective student interaction and learning.

The remainder of this paper is organized as follows. First, we present a concise literature review that discusses the study's theoretical frame and the relationship between purposeful interpersonal interaction and group membership. Then, the student attributes that may be associated with such interaction are discussed, and the associated hypotheses are specified. Following this, we discuss the relationship between purposeful interpersonal interaction and perceived learning, and specify our final hypothesis. After explaining our method, we present the results. We finish the paper with a discussion and conclusion that summarizes and analyzes the key findings, and which explains the study's contributions.

#### Literature review and research hypotheses

#### Theoretical frame

Deep and meaningful learning is possible when high quality student-student interaction occurs (Mehall, 2021). Anderson's (2003) interaction equivalency theory suggests that if student-student interaction is of a high quality, then student-lecturer and student-content interactions may be offered at minimal levels, or even not at all, without negatively affecting students' learning. Online delivery increases the physical distance between the student and lecturer, which may have negative effects on interaction and learning, but Moore's (1983) theory of transactional distance claims that well-designed activities that engage students and effective two-way communication between students and lecturers may decrease the psychological distance between them. Thus, the tasks and activities that

students are asked to perform in breakout rooms may influence the perceived psychological distance between students and lecturers, and individuals' perceived learning and satisfaction with the online learning experience.

According to cognitive load theory, due to their limited working memory, learners have limited cognitive processing capacity to apply acquired knowledge and skills to new situations (Kirschner *et al.*, 2018). If simultaneous speaking and writing (e.g., text chat) are undertaken in the breakout room, higher cognitive load and ambiguity may result, leading to lower quality learning. Levels of interaction, learning, attainment, and satisfaction are influenced by the individual's personality type, e.g., as explained by the Big Five Personality Trait Model (Costa and McCrae, 1992). Although the traits of conscientiousness, openness, agreeableness, and neuroticism may each contribute to explaining an individual's suitability or non-suitability for collaborative learning in a breakout room, extroverts are typically talkative and energetic, and they enjoy activities undertaken in the company of others, making them ideal breakout room participants (Eftekhar *et al.*, 2014). Each of the theories mentioned in this section has implications for the quality of purposeful interpersonal interaction that may occur in online breakout rooms.

#### Purposeful interpersonal interaction and group membership

The main aim of breakout rooms is to achieve purposeful interpersonal interaction between students that will result in effective learning. Purposeful interpersonal interaction may be defined as, 'any high quality, organic, and valid communication exchange between two or more participants of the learning process that directly relates to the achievement of established learning outcomes or to the building of social relationships' (Mehall, 2020, 185). Student-to-student interaction may be critical to students' learning because, as Anderson's (2003) interaction equivalency theory argues, if student-student interaction is of a high quality, then other types of interaction may be offered at minimal levels, or even not at all, without negatively affecting students' learning. With many universities having very large classes, it is often difficult to achieve high quality student-lecturer interaction.

An effective breakout group typically requires individual members to assume different roles, taking responsibility for things such as organization, providing information, creating an action plan, solving problems, motivating others, supporting others, and monitoring/evaluating performance (Yeh, 2010). Role assignment and group size may each influence individual participation, peer interaction, and learning achievement (Luo et al., 2023). A number of studies have concluded that purposeful interaction and collaborative learning is more effective with heterogeneous groups, where there is a mix of students in terms of gender, ethnic background, ability, and personality traits (e.g., Jong et al., 2006; Scheurell, 2010; Wang et al., 2007). In contrast, Chen and Kuo (2019) recommend creating groups where students are heterogeneous in terms of knowledge and learning roles, but homogenous in terms of social interactions among group members. Regardless of whether a student group is heterogeneous or homogenous, individuals who possess an outgoing personality, who can relate well to others, and who have good communication skills may be more willing to interact with their peers in breakout rooms. However, for breakout rooms to be effective, individuals need to have a positive attitude toward online and collaborative learning, and possess technology readiness.

Previous research has indicated that online learning is most effective when the students are ready for online learning, and in particular when they possess technology readiness (Tang *et al.*, 2021). Technological readiness is an essential aspect of student readiness for online learning. For example, Walia *et al.* (2019) specified student readiness for online learning as comprising access to technology, technology skills, teaching presence, cognitive presence, social presence, lifestyle factors, and individual skills and study habits. According to Parasuraman (2000), technology readiness refers to an individual's propensity to embrace and use new technologies for accomplishing specific goals. Furthermore, students' perceived usefulness of online learning may influence their willingness to engage with it (Singh and Tewari, 2021). Warden *et al.* (2020) found that although smart devices are ubiquitous, some students are anxious about online learning, and that students who are less comfortable with technology possess lower self-efficacy in social interactions with their peers. Chau *et al.* (2020), Ferrer *et al.* (2022), and Flores *et al.* (2022) conclude that student technology readiness is critical to achieving effective online learning, as technology readiness promotes a favorable attitude toward online learning, which impacts upon an individual's motivation and engagement.

H1. Students' technology readiness is positively related to purposeful interpersonal interaction in breakout rooms.

#### **Extroversion**

An extrovert enjoys high levels of activity and arousal, and has a tendency toward social behavior, assertiveness, impulsiveness, and the display of positive emotions (Busato *et al.*, 2000). Extroverts find social situations more pleasant than introverts, and therefore are likely to enjoy breakout room interaction with peers more than the introverts. Previous research has found that extroverts are more willing to speak in public and that they experience lower levels of anxiety (Fraj-Andrés *et al.*, 2018). In a study of online learning during the COVID-19 pandemic, Hong et al. (2021) concluded that extroversion can positively predict internet and academic self-efficacy. Although Sanudin *et al.* (2022) found that introverts view online learning positively, these individuals may lurk and not participate in discussion or interaction with peers. Indeed, extrovert students are more likely to engage in purposeful interpersonal interaction in breakout rooms. A study by Weiser *et al.* (2018) found that extroverts spoke more in almost all types of teaching-learning interactions, while Barnett *et al.* (2015) found that extroverts perform better in groups and on tasks requiring significant interaction with others. Extroverts are more likely to be leaders, and leaders are more likely to engage in purposeful interpersonal interaction, which leads to higher academic performance (Dunbar *et al.*, 2018).

*H2*. Students' extroversion is positively related to purposeful interpersonal interaction in breakout rooms.

#### Social identification

Effective learning in breakout rooms requires social interaction between students. When students are allowed to form the breakout groups, they typically choose classmates that are similar to themselves and with whom they have prior acquaintance (Chen and Kuo, 2019). Hilton and Phillips (2010) found that homogenous student-selected groups communicate better and are more enthusiastic about

working together. However, if the lecturer forms heterogeneous groups with dissimilar students, purposeful interpersonal interaction may be lower. The concept of social identification explains that individuals view themselves as belonging to in or out-groups, which results in an 'us' and 'them' mentality. Social identification exists when an individual exhibits common characteristics or behaviors with other members of the in-group. Dean and Jolly (2012) argue that it is a student's sense of identity that determines their level of commitment and willingness to engage with a given learning opportunity. Social identification increases social rapport, which may decrease psychological distance and increase purposeful interpersonal interaction (Woolcott, 1996). Furthermore, Wilkins *et al.* (2016) found that students' social identification is significantly related to students' commitment and achievement.

*H3*. Students' social identification is positively related to purposeful interpersonal interaction in breakout rooms.

#### **Intercultural communication competence**

Today's higher education classrooms typically have students with diverse social, cultural, ethnic, and religious backgrounds. To operate effectively in culturally diverse settings, students need to possess intercultural competence, i.e., the ability to communicate effectively and appropriately in intercultural situations (de Hei *et al.*, 2020). Specifically, intercultural communication competence may be regarded as an individual's ability to achieve their communication goal while effectively and appropriately utilizing communication behaviors to negotiate between the different identities present within the culturally diverse classroom (Portalla and Chen, 2010). Chen and Starosta (1996) conceptualize intercultural communication competence as comprising of intercultural awareness, sensitivity, and effectiveness. Individuals who demonstrate understanding and respect for another's culture and communicate with them using appropriate language and style – e.g., appropriate speed and clarity of speech – are likely to achieve higher quality purposeful interaction (Portalla and Chen, 2010).

*H4.* Students' intercultural communication competence is positively related to purposeful interpersonal interaction in breakout rooms.

#### Perceived learning

Educators expect that breakout rooms generate purposeful interpersonal interaction, and that the purposeful interpersonal interaction results in enhanced student learning. In education research, learning is often measured by students' self-reported perceived learning. Perceived learning refers to students' retrospective evaluation, which results in a set of beliefs and feelings that learning has occurred (Barzilai and Blau, 2014). A number of studies have found a positive relationship between purposeful interpersonal interaction and students' perceived learning (e.g., Blasco-Arcas *et al.*, 2013; Hernández-Sellés *et al.*, 2019; Ku *et al.*, 2013; Mehall, 2020, 2021). Specifically, with regard to breakout rooms, Tonsmann (2014) found that students may develop and apply their understanding of concepts that have been taught by the lecturer in the main room, and he concluded that breakout rooms are a valuable technique to achieve student understanding and assimilation of concepts.

H5. Purposeful interpersonal interaction is positively related to students' perceived learning.

Figure 1 presents the proposed conceptual model and the associated hypotheses.

Insert Figure 1 here

#### Method

#### Sample and data collection

The study population is full-time higher education students who study at institutions based in the United States and who have participated in online lessons in the last 12 months. Data were collected using an online survey questionnaire, which was available through CloudResearch and Prolific. Students are readily available on such platforms. Participant quality control measures were employed (see Litman *et al.*, 2017). Nothing was observed in the participants' answers to suggest that providing compensation for participation generated unreliable or unbelievable responses. A total of 664 usable responses were obtained. Of these respondents, 57.1% classified themselves as male and 42.2% as female. About a quarter of our respondents studied a business/management-related subject; the remainder studied a diverse range of subjects including computer science/information technology, engineering, liberal arts, psychology, and the natural sciences. Average online learning experience in the sample was 1.58 years, and 58% of the students typically participate in at least one breakout room every week; 38% have at least one per month; and remaining 4% participate in less than one per month.

#### Measures

All of the latent constructs in the model were measured using pre-existing validated scales adopted from the literature. All these scales had previously demonstrated satisfactory internal reliability, with Cronbach's alpha scores ranging from .74 to .90 in the original studies. Technology readiness was measured using Tang *et al.*'s (2021) seven item scale reflecting students' ability to acquire and use technologies in life. A six-item extroversion scale was adopted from Francis *et al.* (1992). Wilkins *et al.*'s (2016) five item scale was used to measure students' perceived social identification. The intercultural communication competence scale was adopted from Portalla and Chen (2010), which measures the interaction relaxation and message skills dimensions. An eight item scale adapted from Roblyer and Wiencke (2004) was used to measure purposeful interpersonal interaction. Finally, perceived learning was measured using a four item scale provided by Barzilai and Blau (2014). All latent constructs in this study were measured using a 7-point Likert scale, where 1 = strongly disagree and 7 = strongly agree. The survey questionnaire was subjected to pretesting with 20 students, and no issues of concern were identified. Appendix 1 specifies all of the items used to measure each construct in the model.

#### Data analysis

SPSS version 28 was used to clean the data and assess the demographics. Further analyses were conducted using partial least squares structural equation modeling (PLS-SEM) and the SmartPLS version 4.0.7.8 software, which enabled testing of the hypothesized relationships, and assessment of the predictive power (Sarstedt and Cheah, 2019). A major advantage of using the Smart PLS software is that in addition to providing causal explanation, it also has predictive abilities (Hair *et al.*, 2020), allowing researchers to test the explanatory and predictive power of their models (Shmueli *et al.*, 2019). Although our hypotheses are grounded in causal explanations, we expect our model to have high predictive accuracy and yield meaningful implications for practitioners.

#### **Results**

#### Preliminary data analysis

As the study employed a cross-sectional design, it is important to test if there is any common method bias in the data set. The variance inflation factor (VIF) scores are lower than 3.3, indicating that common method bias is unlikely to be a problem with our data (Kock, 2015). As recommended by Hair *et al.* (2020), we applied confirmatory composite analysis to test the measurement model. As presented in Table 1, all measures for internal consistency – Cronbach's alpha, rho\_A and composite reliability (CR) – are above the cut criteria of 0.7, thus indicating the reliability of our measurement scales (Hair et al. 2019). Also, the scales demonstrated adequate convergent validity, as the average variance extracted (AVE) is higher than .50 for all the constructs. Discriminant validity was assessed using the heterotrait-monotrait (HTMT) ratio of correlations.

Insert Table 1 here

Table 2 presents the results of the test establishing discriminant validity, as all the HTMT values are lower than the conservative threshold value of 0.85 (Henseler *et al.*, 2015). Before testing the structural model, it was first established that no multicollinearity exists between the independent variables in our model. The VIF values range from 1.00 to 1.96, indicating that there is no collinearity between predictors. The relationships between different constructs was then assessed using the bootstrapping procedure with 5,000 subsamples (Hair *et al.*, 2019).

Insert Table 2 here

#### **Hypothesis testing**

Figure 2 presents an overview of the structural model results. All the hypothesized paths are significant except the path from extroversion to purposeful interaction.

### Insert Figure 2 here

The  $R^2$  for purposeful interaction is 61%, (p < .01), and for perceived learning it is 70% (p < .01) (Table 3). All  $f^2$  values are significant except for the relationship between extroversion and purposeful interaction. The effect size for the proposed relationships in H1 and H3 are small, as the values of  $f^2$  are in a range between .11 to .14, while H4 has a medium sized effect and H5 has a large effect size (Cohen, 2003). The blindfolding procedure was used to test the predictive relevance of the structural model. The  $Q^2$  value for purposeful interaction is .37, and for perceived learning it is .44. This establishes that our model has strong predictive relevance, as the test values are non-zero (Chin *et al.*, 2020). These results allow us to assume and test the predictive relevance of the model using PLS predict. If all indicators in the PLS-SEM analysis have lower RMSE (or MAE) values compared to the naïve LM benchmark, the model has high predictive power (Shmueli *et al.*, 2019). Our results indicate that RMSE has lower values for the PLS model as compared to the naïve LM benchmark, and thus our model has high predictive power (Table 4).

Insert Table 3 here

Insert Table 4 here

#### Discussion

The purpose of this research is to identify specific learner attributes that are associated with purposeful interpersonal interaction in breakout rooms. Through answering our research questions, we fill a literature gap that connects the group formation process - i.e., the allocation of students to groups based on their individual attributes - with the use of online breakout rooms, and provide much needed information about the types of student who are more likely to willingly engage in purposeful student-student interaction.

To answer our first research question about what learner attributes are associated with purposeful interpersonal interaction in online breakout rooms, hypotheses 1, 3, and 4 are supported, meaning that students' technology readiness, social identification, and intercultural communication competence are each significantly related to the achievement of purposeful interpersonal interaction. Although our intercultural communication competence construct is concerned mainly with the individual's ability to achieve their communication goal, it also includes the behaviors needed to negotiate between the different identities present within the culturally diverse classroom. In other words, intercultural communication competence is not just about the spoken words, but also about intercultural awareness and sensitivity.

The relationship between intercultural communication competence and purposeful interpersonal interaction has the largest effect size among our four learner attributes. Thus, our findings support the existing literature that identifies cultural differences and English language competence as key factors which influence the success of students in culturally diverse settings (Straker, 2016). Universities should aim to develop students' intercultural competence both in and out of the classroom. To increase understanding and familiarness among students, lecturers may use breakout rooms for icebreaker activities at the start of a course, when students are still getting to know one another. Students could share photos and other personal media artifacts to explain their backgrounds and interests in ways that allow their fellow community of learners to comment and engage in conversations around common interests and experiences.

Extracurricular activities, such as sports and special interest groups, may be a vehicle for developing students' social identification. Such events may promote interaction, understanding, and familiarity between students. To ensure that students have the technical competence to be confident and effective in online lessons, universities should offer appropriate training on using the online learning platform and associated learning resources. Also, if the lecturer ensures that each group has at least one individual who is technically competent, these individuals may offer support to other students who are less confident with using the technology.

For our second research question, about the relationship between purposeful interpersonal interaction and students' perceived learning (H5), it was found that the relationship is significant and with a large effect size. This was to be expected, as it is a logical relationship that is well documented in the literature (e.g., Blasco-Arcas *et al.*, 2013; Hernández-Sellés *et al.*, 2019; Ku *et al.*, 2013; Mehall, 2020, 2021). The large effect size for the relationship between purposeful interpersonal interaction and students' perceived learning emphasizes the value of group discussion and activities to the individual's learning. It is important that lecturers create group tasks and activities that stimulate and engage students, and which require effective team working. Lecturers may, for example, use graphic organizers and apps – such as Nearpod, Pear Deck and Jamboard – to make breakout rooms more engaging.

#### Conclusion

The findings of this research emphasize the importance of lecturers considering learner attributes when forming breakout room groups. To our knowledge, this is the first study to investigate the relationships between a selection of learner attributes and purposeful interaction for effective collaborative learning in online breakout rooms. Thus, this research fills an important literature gap. We propose and validate a conceptual model for purposeful interpersonal interaction and learning in online lessons, specifically in the context of breakout rooms. Thus, the theoretical contribution is specific to student interaction and learning in breakout rooms, which represent a quite unique learning environment, where students are expected to interact and work independently with minimal or no lecturer input and where they can easily disengage by switching off their cameras and/or microphones. Our model identifies three learner attributes that may be associated with purposeful interpersonal interaction and effective learning.

Lecturers need to identify individuals' prior experience and training needs before sending students into a breakout room. They must also be aware that students with different social, cultural, and ethnic backgrounds will likely have different individual attributes and different attitudes toward online collaborative learning. It is important for lecturers to monitor students' performance in breakout rooms by visiting each room as much as possible, to provide advice, support, and feedback to individuals and groups, and not to use the time that students are in breakout rooms to take a rest or catch up with other work tasks. The implications of our findings for practice and our recommendations for educators are as follows. First, lecturers need to be organized and pre-plan their online lessons. Second, lecturers need to design learning tasks that will engage students and which enable individuals to assume a specific group role for which they are suitable. Third, lecturers must recognize that students have different characteristics, attributes, and learning preferences and for any particular learning task lecturers need to determine whether a homogeneous or heterogeneous group would be more effective as well as the optimal group size to maximize individual participation and interaction. This need implies that lecturers should have access to personal information about their students, and that lecturers should make every effort to become familiar with their students so that they can identify individuals' key characteristics and attributes.

As with all research, there are some limitations to acknowledge. The study adopted a cross-sectional research design and data were collected only in one country. Future studies might use a longitudinal design to capture the students' development in technological competence, social identification, and intercultural competence, to assess the impacts on students' purposeful interaction and learning. It would be interesting to discover whether our results are replicated in different countries and with groups of learners with different social, cultural, and ethnic backgrounds. Our study considered only four antecedents of purposeful interaction, but future studies could add further predictor variables, like student self-efficacy. It would also be interesting in future to explore the impacts of the lecturer's characteristics and task design on the students' purposeful interactions in online breakout rooms.

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Insert Appendix 1 here

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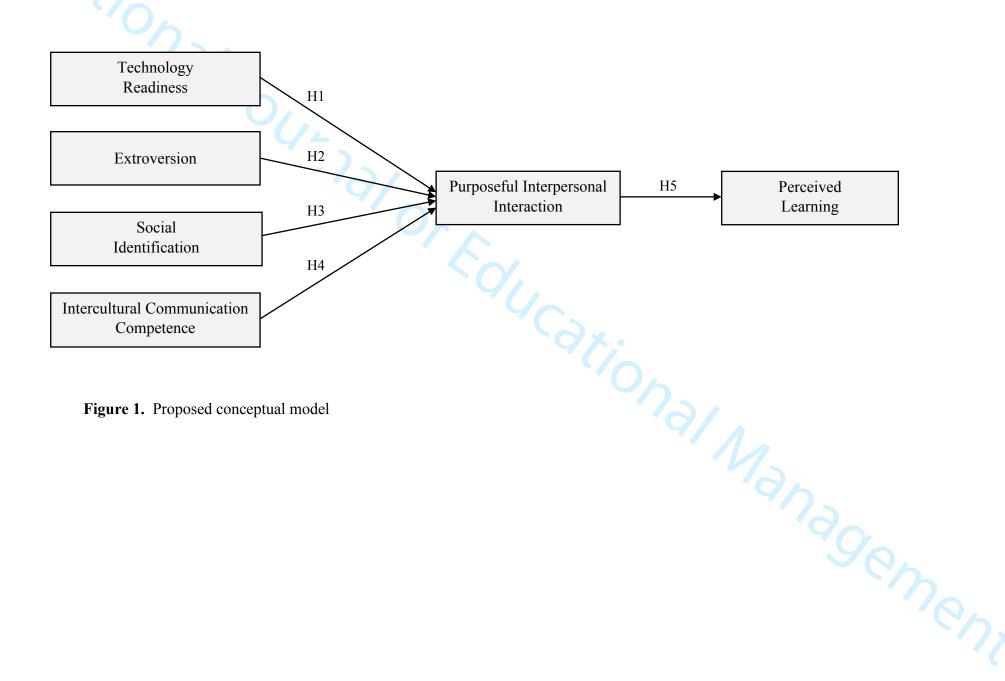


Figure 1. Proposed conceptual model

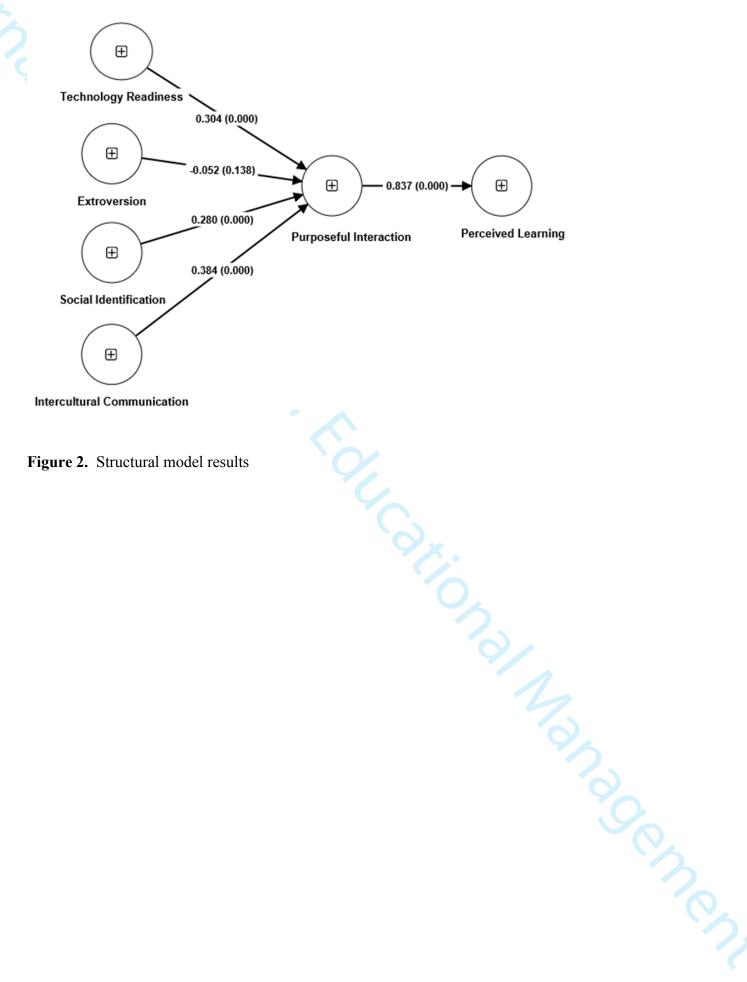


Figure 2. Structural model results

Table 1. Assessment onloading, full collinearity, reliability, and convergent validity

TR1 TR2 TR3 TR4 TR5 TR6 TR7 EX1 EX2 EX3 EX4	0.772 0.733 0.773 0.678 0.762 0.717 0.734 0.815 0.849 0.849	38.25 34.86 34.01 26.11 40.02 27.83 31.50 47.16 68.25	0.73; 0.80 0.68; 0.77 0.72; 0.81 0.62; 0.73 0.72; 0.80 0.66; 0.77 0.68; 0.77 0.78; 0.85	1.95	0.89	0.862	0.894
TR3 TR4 TR5 TR6 TR7 EX1 EX2 EX3	0.773 0.678 0.762 0.717 0.734 0.815 0.849	34.01 26.11 40.02 27.83 31.50 47.16	0.72; 0.81 0.62; 0.73 0.72; 0.80 0.66; 0.77 0.68; 0.77 0.78; 0.85	1.73	0.90		
TR4 TR5 TR6 TR7 EX1 EX2 EX3	0.678 0.762 0.717 0.734 0.815 0.849	26.11 40.02 27.83 31.50 47.16	0.62; 0.73 0.72; 0.80 0.66; 0.77 0.68; 0.77 0.78; 0.85	1.73	0.90		
TR5 TR6 TR7 EX1 EX2 EX3	0.762 0.717 0.734 0.815 0.849	40.02 27.83 31.50 47.16	0.72; 0.80 0.66; 0.77 0.68; 0.77 0.78; 0.85	1.73	0.90		
TR6 TR7 EX1 EX2 EX3	0.717 0.734 0.815 0.849	27.83 31.50 47.16	0.66; 0.77 0.68; 0.77 0.78; 0.85	1.73	0.90		
TR7 EX1 EX2 EX3	0.734 0.815 0.849	31.50 47.16	0.68; 0.77 0.78; 0.85	1.73	0.90		
EX1 EX2 EX3	0.815 0.849	47.16	0.68; 0.77 0.78; 0.85	1.73	0.90		
EX2 EX3	0.849	47.16		1.73	0.90		
EX3		68.25			0.70	0.914	0.930
EX3			0.82; 0.87				
	U.UT/	69.21	0.82; 0.87				
	0.812	55.62	0.78; 0.84				
EX5	0.758	34.59	0.71; 0.80				
EX6	0.887	86.03	0.87; 0.91				
				2.11	0.86	0.885	0.916
				2 24	0.89	0.895	0.914
					0.05	0.050	0.71.
			•				
				2 60	0.90	0.902	0.931
				2.00	0.50	0.702	0.731
				2.51	0.88	0 008	0.925
				2.31	0.00	0.700	0.943
	SI1 SI2 SI3 SI4 SI5 ICC1 ICC2 ICC3 ICC4 ICC5 ICC6 ICC7 ICC8 PII1 PII2 PII3 PII4 PII5 PII6 PII7 PII8 PL1 PL2 PL3 PL4	SI2 0.843 SI3 0.846 SI4 0.819 SI5 0.805 ICC1 0.793 ICC2 0.789 ICC3 0.781 ICC4 0.770 ICC5 0.810 ICC6 0.665 ICC7 0.676 ICC8 0.747 PII1 0.800 PII2 0.786 PII3 0.770 PII4 0.808 PII5 0.766 PII6 0.811 PII7 0.747 PII8 0.740 PL1 0.868 PL2 0.878 PL3 0.892	SI2       0.843       58.63         SI3       0.846       55.26         SI4       0.819       53.96         SI5       0.805       44.55         ICC1       0.793       43.11         ICC2       0.789       40.78         ICC3       0.781       48.84         ICC4       0.770       34.82         ICC5       0.810       49.56         ICC6       0.665       22.49         ICC7       0.676       23.67         ICC8       0.747       32.38         PII1       0.800       48.33         PII2       0.786       42.61         PII3       0.770       38.65         PII4       0.808       51.67         PII5       0.766       40.04         PII6       0.811       53.18         PII7       0.747       31.60         PII8       0.740       36.25         PL1       0.868       62.14         PL2       0.878       76.91         PL3       0.892       82.95	SI2       0.843       58.63       0.81; 0.87         SI3       0.846       55.26       0.81; 0.87         SI4       0.819       53.96       0.78; 0.84         SI5       0.805       44.55       0.76; 0.83         ICC1       0.793       43.11       0.75; 0.82         ICC2       0.789       40.78       0.75; 0.82         ICC3       0.781       48.84       0.75; 0.81         ICC4       0.770       34.82       0.72; 0.81         ICC5       0.810       49.56       0.77; 0.84         ICC6       0.665       22.49       0.60; 0.72         ICC7       0.676       23.67       0.62; 0.73         ICC8       0.747       32.38       0.70; 0.79         PII1       0.800       48.33       0.76; 0.83         PII2       0.786       42.61       0.75; 0.82         PII3       0.770       38.65       0.73; 0.80         PII4       0.808       51.67       0.78; 0.83         PII5       0.766       40.04       0.72; 0.80         PII6       0.811       53.18       0.78; 0.84         PII7       0.747       31.60       0.70; 0.79      <	SI2	SI2	SI2

Table 2. Heterotrait-Monotrait Ratio (HTMT) results

- 0.59 0.66 0.62 0.73 0.61	- 0.62 0.62 0.53 0.50	- 0.67 0.73 0.65	- 0.77 0.58	Interaction  - 0.84	-
0.66 0.62 0.73	0.62 0.53 0.50	0.73 0.65	0.58		-
0.62	0.62 0.53 0.50	0.73 0.65	0.58		-
0.73	0.53	0.73 0.65	0.58		-
	0.50	0.65	0.58		-
0.61		),			-
		740			

Table 3. Structural model results

Path	Coefficient	SD	T Value	Confi	idence	$f^2$	R <sup>2</sup>	$Q^2$	SRMR	
				Inte	rvals	· ·		~		
				2.50%	97.50%					
H1 TR to PII	0.30	0.03	8.01*	0.23	0.38	.141*				
H2 EX to PII	-0.05	0.03	1.48	-0.12	0.17	.004				
H3 SI to PII	0.28	0.04	6.39*	0.19	0.36	.107*				
H4 ICC to PII	0.38	0.04	8.42*	0.29	0.47	.208*	.61	.37		
H5 PII to PL	0.84	0.02	55.53*	0.80	0.86	2.370*	.70	.44	0.053	
<i>p</i> < 0.01										

<sup>\*</sup>p < 0.01

Table 4. PLS predict results

	90	PLS Model			LM (Linear			PLS-LM		
	RMSE	MAE	Q <sup>2</sup> predict	RMSE	model) MAE	Q <sup>2</sup> _predict	RMSE	MAE	Q <sup>2</sup> predict	
PII1	0.963	0.716	0.361	1.004	0.746	0.306	-0.041	-0.030	0.055	
PII2	0.918	0.689	0.359	0.949	0.710	0.316	-0.031	-0.021	0.043	
PII3	0.961	0.738	0.348	0.992	0.761	0.305	-0.031	-0.023	0.043	
PII4	0.914	0.686	0.463	0.948	0.710	0.422	-0.034	-0.024	0.041	
PII5	1.023	0.765	0.324	1.063	0.794	0.270	-0.040	-0.029	0.054	
PII6	1.002	0.734	0.367	1.034	0.768	0.327	-0.032	-0.034	0.040	
PII7	1.003	0.750	0.366	1.010	0.769	0.356	-0.007	-0.019	0.010	
PII8	1.013	0.771	0.354	1.046	0.795	0.311	-0.033	-0.024	0.043	
PL1	1.236	0.921	0.279	1.222	0.903	0.296	0.014	0.018	-0.017	
PL2	1.085	0.806	0.326	1.096	0.810	0.312	-0.011	-0.004	0.014	
PL3	1.120	0.815	0.315	1.135	0.836	0.296	-0.015	-0.021	0.019	
PL4	1.142	0.831	0.314	1.161	0.847	0.291	-0.019	-0.016	0.023	

Appendix 1. Measurement scales used in this study

Construct	Items	Source	α in original study
Technology	TR1 - I prefer to use the most advanced technology available	Tang et al. (2021)	.75
Readiness	TR2 - Technology gives me more freedom of mobility		
(TR)	TR3 - I feel confident that machines will do what I have instructed them to do		
	TR4 - I am usually among the first to acquire new technology when it appears		
	TR5 - I enjoy the challenge of understanding high-tech gadgets TR6 - Technology is usually reliable		
	TR7 - Technology transmits information effectively to the intended recipient		
Extroversion	EX1 - I am a talkative person	Francis, Brown, &	Between
(EX)	EX2 - I am a lively person	Philipchalk, R. (1992)	.74 and
(LA)	EX3 - I can usually bring some life into a dull situation	1 IIIIpenaik, R. (1992)	.84
	EX4 - I am rarely in the background in social occasions		.04
	EX5 - I am not quiet when I am with other people		
	EX6 - Other people think that I am lively		
Social	SII - I feel a bond with the other students in my degree program	Wilkins et al. (2016)	.88
Identification	SI2 - It is pleasant to be a member of the student cohort in my degree program	(2010)	.00
(SI)	SI3 - Being a member of the student cohort in my degree program gives me a good		
	feeling		
	SI4 - Fellow students are a source of friendship for me		
	SI5 - Fellow students are a source of future networking for me		
Intercultural	ICC1 - I find it easy to talk with people from different cultures	Portalla & Chen (2010)	.87
Communication	ICC2 - I can always initiate a conversation with people from different cultures	, ,	
Competence	ICC3 - I find it is easy to identify with students from different cultures		
(ICC)	ICC4 - I find it is easy to get along with students from different cultures		
	ICC5 - I feel relaxed when interacting with students from different cultures		
	ICC6 - I don't have grammar problems when interacting with students from different		
	cultures		
	ICC7 - There are rarely misunderstandings when I interact with students from		
	different cultures		
	ICC8 - I usually understand messages from students with a different culture to me		
Purposeful	PII1 - Communication and relationships between students are positive in breakout	Roblyer & Wiencke	.85
Interpersonal	rooms	(2004)	(average)
Interaction	PII2 - Communication and relationships between students and the instructor are		
(PII)	positive		
	PII3 - Students work together cooperatively in groups		

PL4 - I remember things I learn in breakout rooms

Perceived

Learning (PL)

PII4 - Technologies are used effectively in online group work for two-way exchanges of information PII5 - Most students initiate and reply to messages from other students PII6 - Communication between students in online group work is detailed and beneficial to learning PII7 - Instructors provide useful feedback on students' online group work PII8 - Instructors respond to student queries promptly during online student group working arn in breakout rooms

.s

.arn in breakout rooms PL1 - I learn a lot in breakout rooms Barzilai & Blau (2014) .90 PL2 - I gain new knowledge when working with other students in online groups PL3 - I learn new things when working with other students in online groups