The relationship between user types and gamification designs

Ana Cláudia Guimarães Santos1 · Wilk Oliveira1 · Juho Hamari2 · Luiz Rodrigues1 · Armando M. Toda1 · Paula T. Palomino1 · Seiji Isotani1

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Abstract
Gamification has been discussed as a standout approach to improve user experience, with different studies showing that users can have different preferences over game elements according to their user types. However, relatively less is known how different kinds of users may react to different types of gamification. Therefore, in this study (N = 331) we investigate how user orientation (Achiever, Disruptor, Free Spirit, Philanthropist, Player, and Socializer) is associated with the preference for and perceived sense of accomplishment from different gamification designs. Beyond singular associations between the user orientation and the gamification designs, the findings indicate no comprehensive and consistent patterns of associations. From the six user orientations, five presented significant associations: Socializer orientation was positively associated with Social, Fictional, and Personal designs, while negatively associated with Performance design; Player orientation was positively associated with Social (Accomplishment), Personal, and Ecological designs, while negatively associated with the Social design (Preference); Disruptor orientation was positively associated with Social design; Achiever orientation was positively associated with Performance and Social designs; and Free Spirit orientation was negatively associated with Social design. Based on the results, we provide recommendations on how to personalize gamified systems and set further research trajectories on personalized gamification.

Keywords User modeling · Personalization · Gamification · Hexad · Gameful experience

Ana Cláudia Guimarães Santos
anaclaudiaguimaraes@usp.br

Juho Hamari
juho.hamari@tuni.fi

1 Institute of Mathematics and Computer Science, University of São Paulo, São Carlos, São Paulo, Brazil
2 Gamification Group, Faculty of Information Technology and Communication Sciences, Tampere University, Tampere, Finland
1 Introduction

Gamification has been widely used in recent years to increase users’ motivation in different areas, such as health (Johnson et al. 2016), virtual reality (Hassan et al. 2020), and education (Barata et al. 2013; Turan et al. 2016; Araya et al. 2019). According to Koivisto and Hamari (2019), the education/learning context is the most common in studies about gamification, and most have reported positive results. One goal of the use of gamification in education is to lead students to desired psychological outcomes (e.g., engagement, motivation, fun, or autonomy (Majuri et al. 2018)); however, some studies have also reported that gamification can have negative effects on students’ behavior (Hanus and Fox 2015; Toda et al. 2017; Bai et al. 2020).

One of the main hypotheses for these negative effects is that people have different player types (i.e., different characteristics and preferences over game elements (Nacke et al. 2014)), which leads to different perceptions regarding the gamification design (Lavoué et al. 2018; Hallifax et al. 2019b; Oliveira et al. 2020), that can positively or negatively be affected by some of the game’s elements (Oliveira and Bittencourt 2019b). At the same time, most of the gamified systems are developed in a way called “one size fits all”, which means that the users’ preferences are ignored and normally the designers create a universal gamified environment to suit all users (Tondello et al. 2017b; Oliveira et al. 2018), thus possibly negatively affecting their experience (Tondello et al. 2017b; Lavoué et al. 2018; Rodrigues et al. 2019).

Although in the past few years researchers have conducted some studies about personalized gamification (Hallifax et al. 2019a; Klock et al. 2020; Rodrigues et al. 2020), they have not reached a consensus about which game elements would be the most suitable for each player/user type, have used a small number of game elements, or have analyzed the relation of user types and game elements individually (Hallifax et al. 2019b; Klock et al. 2020). The relation of user types and sets of game elements demonstrated in the study conducted by Tondello et al. (2016) showed that the user types could be related with sets of game elements, rather than individual game elements. When we consider the way that the game elements are selected, most of the studies about personalized gamification select the game elements deliberately or by using literature reviews, which can bring some limitations (e.g., the use and random naming of game elements that are correlated, or the exclusion of a game element that would be suitable for the context) (Klock et al. 2020; Rodrigues et al. 2020). Thus, one gap in the personalized gamification of gamified systems is studies about the preference for gamification designs (sets of game elements grouped according to their characteristics or purpose) for each user type and if they are positively affected by these gamification designs.

We tackled this challenge through a study with 331 participants, where we (1) identified their user types, (2) analyzed their preferences regarding different gamification designs (represented in storyboards), (3) measured the participants’ preference and perceived sense of accomplishment in each gamification design,
and (4) analyzed the participants’ preference and perceived sense of accomplishment according to their user orientations, thus advancing toward answering the question: **How are user types associated with preference and perceived sense of accomplishment in different gamification designs?** Our results allow us to move toward evidence-based gamification design, generating new insights for gamification designers to create more effective gamified systems according to the users’ preferences and experiences. We also provided a series of validated storyboards to represent the design concept of personalized gamified educational system.

This article is structured as follows: In Sect. 2, we present the study background, including an overview of player/user typologies, the gamification taxonomy used to select game elements, the gameful experience and why we measured one of its dimensions, and the main related work. Following on, we describe in Sect. 3 how the study was conducted, the construction and validation of the storyboards to represent the gamification designs, the survey’s construction and application, and the main characteristics of the participants. In Sect. 4, we present and discuss the data collected and our statistical results. We also present the limitations of our study, as well as recommendations for future studies. Finally, in Sect. 5 we present the final remarks of our study.

## 2 Background

This section presents our study background (i.e., player/user types, gamification taxonomies, and gameful experience), as well as the main related works.

### 2.1 Player/user types

Throughout the years, researchers have worked on how certain characteristics may affect the user’s engagement while using a gamified system (Ferro et al. 2013) and how people can be grouped into player types (Yee 2006; Nacke et al. 2011). One of the first player type models was presented by Bartle (1996), which proposed a classification of four player types: (1) Achiever; (2) Explorer; (3) Killer; and (4) Socializer. Based on Bartle’s player types, Yee (2006) proposed an empirical model of player motivations, based on data collected from 3000 Massive Multiplayer Online Role Playing Games (MMORPGs). In his analysis, Yee revealed ten motivation sub-components (Advancement, Mechanics, Competition, Socializing, Relationship, Teamwork, Discovery, Role-Playing, Customization, and Escapism), which he grouped into three overarching components (Achievement, Immersion and Social).

Another player type model that has been used in researches is the BrainHex Model (Nacke et al. 2011), which was based on neurobiological findings and has seven player types: (1) Seeker; (2) Survivor; (3) Daredevil; (4) Mastermind; (5) Conqueror; (6) Socializer; and (7) Achiever. According to Nacke et al. (2011), each player type from the BrainHex model should be understood not as a psychometric type, but rather as an archetype that typifies a particular player experience.
To create a model designed specifically for gamification, Marczewski (2015) proposed theGamification User Types Hexad, with six user types motivated by intrinsic or extrinsic motivational factors. The user type division in intrinsic and extrinsic motivation is based on the self-determination theory (SDT) that says that people are intrinsically motivated when the activity supports three basic human psychological needs (competence, autonomy and relatedness), or extrinsically motivated when the reason for doing something is not an interest in the activity itself (Deci and Ryan 1985). According to Diamond et al. (2015) and Tondello et al. (2016), the user types motivated by intrinsic motivations are the (1) Socializers; (2) Free Spirits; (3) Achievers; and (4) Philanthropists, while (5) Players are motivated by extrinsic motivations. The (6) Disruptors are not a user type derived from SDT, but from the observation of user behavior within online systems (Tondello et al. 2019).

The Hexad has been chosen for our study since it is considered the most appropriate user typology for tailoring gamification (Hallifax et al. 2019b), it does not classify the user in one specific user type (users are classified in more than one user type, with a principal tendency followed by others in some degree (Tondello et al. 2016)), and the model is empirically validated (Tondello et al. 2019), was created especially for gamification (Marczewski 2015), and has been successfully used in other recent studies (Orji et al. 2018; Lopez and Tucker 2019; Mora et al. 2019; Hallifax et al. 2020).

2.2 Gamification taxonomy

Even though there are several gamification frameworks (Azouz and Lefdaoui 2018), only a few of them are developed for the educational context (Toda et al. 2019). To help designers, teachers and instructors select and understand how game elements can be used in the educational context; Toda et al. (2019) created a gamification taxonomy composed of twenty-one game elements that could be used in gamified educational environments, organizing them in five dimensions. The Performance/Measurement dimension is related to the environment response and has the elements Point, Progression, Level, Stats, and Acknowledgement. The Ecological dimension is related to the environment that the gamification is implemented in, and is formed by the elements Chance, Imposed Choice, Economy, Rarity, and Time Pressure. The Social dimension is related to the interactions between the learners presented in the environment and has the elements Competition, Cooperation, Reputation, and Social Pressure. The Personal dimension is related to the learner that is using the environment and has the elements Sensation, Objective, Puzzle, Novelty, and Renovation. Finally, the Fictional dimension is the mixed dimension that is related to the user and the environment and has the elements Narrative and Storytelling.

As far as we know, this taxonomy is the only one that has been developed and validated for the educational context, explaining a considerable number of game elements, and grouped them into dimensions. Considering that the deliberate selection of game elements can lead to the use of different game elements with the same goal...
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(Rodrigues et al. 2020), we decided to use Toda’s taxonomy to select the game elements for the storyboards’ design for this study.

2.3 Gameful experience

The success of gamification depends on the gameful experience\(^1\) the service creates in the user (Eppmann et al. 2018). Also, new generations seem to be more susceptible to having gameful experiences (Högberg et al. 2019), showing that its measurement can be an important part of the future design of gamified environments.

Högberg et al. (2019) presented a validated instrument for measuring the gameful experience users can have while using a system or a service. They also identified seven dimensions that describe the gameful experience: Accomplishment, Challenge, Competition, Guided, Immersion, Playfulness, and Social experiences. The instrument is formed of 56 questions and can be used for adaptive gamification and user-modeling research in gamification contexts.

The accomplishment dimension is defined as experiencing the demand for successful performance, goal achievement and progress (Högberg et al. 2019). Users can be motivated to complete a goal or task for the pleasure of feeling accomplished, as there is a specific type of intrinsic motivation that leans toward accomplishment (Vallerand et al. 1992; Barkoukis et al. 2008). In our study, we focused on the measurement of this dimension because it can reflect the users’ engagement and can be considered as a long-term experience that extends beyond the use of the service, which can be essential to achieving the goal of gamification (Högberg et al. 2019).

2.4 Related work

Different studies have been conducted to relate users with game elements considering gamification in general, and also in specific domains (Koivisto and Hamari 2019). Moreover, the personalization of gamification has been the object of some literature reviews conducted over recent years (Hallifax et al. 2019a; Klock et al. 2020; Rodrigues et al. 2020). We used these literature reviews to conduct a snowballing review to find studies that have looked at how player/user typologies can be related to game elements in the educational context. In this section, we briefly discuss some of these recent studies.

Using the BrainHex typology, Monterrat et al. (2017) proposed a model and an adaptation process to gamify learning environments to increase learners’ motivation. They tested the adaptation process with 59 middle school students, where they showed that different player types have different preferences and perceptions for game elements. They also showed that increasing the game elements in an environment also increases the perceived complexity of the environment for the students.

\(^1\) In this study, we used the definition proposed by Landers et al. (2019): “A psychological state where the user perceives non-trivial achievable goals created externally, is motivated to pursue them under an arbitrary set of behavioral rules, and evaluates that motivation as voluntary”.

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Lavoué et al. (2018) conducted an experiment with 266 participants, and according to their player types, they adapted the gamification features that were displayed in the interface of an online learning environment. Their results showed that the participants with the adapted features and counter-adapted features had similar values in their enjoyment, although the adapted features improved the participation of learners who used the environment for more time, and reduced the learners’ level of amotivation when compared to counter-adaptive gamification. Oliveira and Bittencourt (2019a) conducted an empirical experiment with 121 elementary students to identify the students’ preferences for game elements according to their BrainHex player type. Corroborating the results of Montrerrat et al. (2017), they also identified that students had different preferences for the game elements according to their player type. They further suggested a guideline to design or adapt gamified educational environments based on the BrainHex player type and ten game elements. Daghestani et al. (2020) used gamification, classification, and adaptation techniques to study the impact of gamification on students’ engagement and learning. From three different groups of students (control group, gamified group, and adaptive-gamification group), they showed that the students who used an adaptive-gamified system presented a better performance than the students who used the gamified system, and that the gamification and adaptive-gamification had a positive effect on students engagement. The adaptive-gamified system was tailored based on ten game elements and BrainHex player types.

One of the first studies to use Hexad to explore the user type was conducted by Gil et al. (2015). Their study used an initial version of the Hexad (with four user types: Achievers, Socializers, Explorers, and Philanthropists) to correlate the user types with 19 game elements and 21 students’ actions in an e-learning environment. They conducted a user study with students from a 1st-year course of Computer Science, where the students could freely choose their actions in the e-learning environment. Their results showed that for all of the user types except Explorer, the students’ actions and game elements were related to the correspondent user types. Tondello et al. (2016) analyzed the correlation between the Hexad user types and 32 game elements, based on the preference of students from the University of Waterloo. Using the Hexad suggestion (Marczewski 2015), they also grouped the 32 game elements into six sets (one for each user type), analyzing the correlation with the correspondent user type. They found positive correlations between almost all the Hexad user types and the proposed game elements, except the Philanthropists. Based on their results, they created a new association table, suggesting a set of game elements for each Hexad user type. Hallifax et al. (2020) conducted a study with students from high schools, tailoring six game elements with their Hexad user types and the initial motivation user model. Considering the results of tailoring gamification based only on the Hexad user types, they found that students were more engaged in the learning task when they used game elements tailored for their user types. Furthermore, this higher level of engagement was associated with lower student performances, and their results showed that an adaptation based only on player types had no effect on learner motivation to learn Mathematics.

Even though our gamification designs demonstrated instructional content in gamified education, we did not limit the educational level of the respondents as most of
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The related works did (Gil et al. 2015; Tondello et al. 2016; Monerrat et al. 2017; Oliveira and Bittencourt 2019a; Daghestani et al. 2020; Hallifax et al. 2020). Also, while all of the studies featured in our related works selected the game elements based on other studies (Monerrat et al. 2017; Lavoué et al. 2018; Daghestani et al. 2020), most used in literature (Oliveira and Bittencourt 2019a; Hallifax et al. 2020), or in the suggestion of the user typology (Gil et al. 2015; Tondello et al. 2016), we selected the game elements based on a gamification taxonomy created specifically for educational environments. Except for the study conducted by Tondello et al. (2016), all of the studies only evaluated the correlation of the user types with the game elements individually. Thus, as far as we know, our study is the first to conduct an empirical study evaluating respondent preference regarding different gamification designs (a group of strategically organized game elements) in educational settings, considering a gamification-based user typology, a validated taxonomy for the education domain, and also a dimension of the gameful experience. In Table 1, we present a comparison between the related works.

3 Study design

Our study aimed to identify whether the user orientations affect the preference and perceived sense of accomplishment for gamification designs in gamified systems. Thus our research question is: “How are user orientations (Philanthropist, Achiever, Socializer, Free Spirit, Player and Disruptor) associated with preference and perceived sense of accomplishment in different gamification designs (Fictional, Personal, Performance, Social and Ecological)?”. To answer our research question, we organized our study in five different steps: (1) storyboards design; (2) survey design; (3) pilot study; (4) survey application; and (5) data analysis. Figure 1 summarizes the study design.

3.1 Materials and method

Before designing our survey, we needed to decide how to present the dimensions of the gamification taxonomy in a way that the respondents could imagine how their implementation would be in a real gamified system. As recommended in recent literature in the field of human-computer interaction (HCI), a good strategy is the use of storyboards (Orji et al. 2014; Altmeyer et al. 2019, 2020; Yassaee et al. 2019). Storyboards are a graphical depiction of a narrative (Truong et al. 2006) that can be used in HCI and design to illustrate interfaces and contexts of use, thus offering designers a possibility as a prototyping technique. The use of storyboards can help users perceive and interpret proposed functionalities (Truong et al. 2006), and also help to direct the respondents’ focus (Yassaee et al. 2019). Given the benefits storyboards offer to collect user reactions to the system elements (Truong et al. 2006), that they can help with the collection of data from people with different backgrounds since they provide a visual language that facilitates user understanding (Orji et al. 2018), and that other recent gamification studies have used storyboards (Altmeyer
### Table 1  Related works comparison

| Study/Year            | Typology | GE  | SE                                      | P    | ME                                           |
|-----------------------|----------|-----|-----------------------------------------|------|----------------------------------------------|
| Gil et al. (2015)     | Hexad    | 19  | Based on Hexad suggestion (Marczewski 2015) | 40   | Students’ actions                            |
| Tondello et al. (2016)| Hexad    | 32  | Based on Hexad suggestion (Marczewski 2015) | 133  | Preference for game elements                 |
| Monterrat et al. (2017)| BrainHex| 3   | Based on literature                     | 59   | Students’ interactions with the game elements and perceived fun and usefulness |
| Lavoué et al. (2018)  | BrainHex | 5   | Based on literature                     | 266  | Time spent in the environment, enjoyment and motivation |
| Oliveira and Bittencourt (2019a) | BrainHex | 10  | Based on most used in literature        | 121  | Preference for game elements                 |
| Daghestani et al. (2020) | BrainHex| 10  | Based on literature                     | 77   | Students’ interaction data and engagement    |
| Hallifax et al. (2020) | Hexad    | 6   | Based on most used in literature        | 258  | Motivation and engagement                    |
| Our study             | Hexad    | 21  | Validated taxonomy for the educational context | 331  | Preference for game elements and perceived sense of accomplishment |

**Typology:** player/user typology; **GE:** amount of game elements used; **SE:** way that the game elements were selected; **P:** number of participants; **ME:** what was evaluated (e.g., preference for game elements; flow, engagement)
et al. 2019; Yassaee et al. 2019; Altmeyer et al. 2020), we decided to implement five storyboards to represent each of the five dimensions proposed in Toda’s taxonomy (Toda et al. 2019).

To design the storyboards, we followed the recommendations of Truong et al. (2006), which have been successfully used in recent similar studies (e.g., Orji et al. (2017); Altmeyer et al. (2019); Brenes et al. (2019)). Truong et al. (2006) determined five attributes to design a storyboard: (i) Level of detail; (ii) Inclusion of text; (iii) Inclusion of people and emotions; (iv) Number of frames; and (v) Portrayal of time. Thus, we created five storyboards with six frames each, representing a fictional learning environment without defining a specific curricular component. All of the 21 game elements of Toda’s taxonomy (Toda et al. 2019) were used in the storyboards according to their dimension. Also, considering the own organization of the taxonomy to avoid overlapping, each one of the 21 game elements was represented only in one storyboard.

To ensure the storyboards’ quality, after the storyboards’ design, they were evaluated by three gamification experts with extensive experience in evaluating this type of technology. Two experts had six years’ experience in researching gamification, and one had nine years. To conduct the evaluation, we used a Likert scale (Likert 1932) that went from 1 (totally disagree) to 5 (totally agree), asking the experts about the storyboards. They had to answer the following questions: (1) “Does this storyboard represent the dimension?”; and (2) “How can we improve this storyboard?” This last was an open question, so they could give their impressions about the storyboard and tell us how to improve it.

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2 The storyboards were designed at: https://www.storyboardthat.com/.
Our main goal in this evaluation was to guarantee that the storyboards correctly represented the five dimensions proposed by Toda et al. (2019). In their evaluation, only one storyboard got an evaluation of 1 (totally disagree) in relation to the question “Does this storyboard represent the dimension?” by one of the experts. In the other storyboards, all of them evaluated as 5 (totally agree) or 4 (partially agree) for the same question. Then, we updated the storyboards according to their feedback to the question “How can we improve this storyboard?”: the gamification experts pointed out some game elements that were present in more than one storyboard, that the use of some figures could cause a negative feeling about a prize, and offered views on how we could improve the use of the game element Competition. The storyboards and their textual description can be seen in the Appendix.

After the evaluation and improvement in the storyboards, we designed the survey. The survey was composed of 71 questions organized in three different sections. (1) Demographic information: gender, age, education degree, and gaming habits. (2) User type identification: we used the Gamification User Types Hexad (Tondello et al. 2016), thus the respondents were asked to rate how well the 24-item scale proposed by Tondello et al. (2016) represented them. We used a 7-point Likert scale (Likert 1932), the questions were presented in a random order, and respondents could not identify the corresponding type (as recommended by Tondello et al. (2016)). Inspired by other studies (Orji et al. 2018; Hallifax et al. 2019b; Oliveira et al. 2020), in this part of the survey we used an “attention-check” question: “I like to be with my friends, but this question is just to evaluate your attention. Please, mark the option number 3, to let us know that you are paying attention”. This question was to ensure that the respondents were paying attention in the survey and reading all the items.

Finally, the last section was the (3) sense of accomplishment and preference measurement. In this step, we used the sub-scale proposed by Högberg et al. (2019) to measure the perceived sense of accomplishment for each gamification design. Following Högberg et al. (2019) recommendations, the respondents were asked to rate on a 7-point Likert scale (Likert 1932) how well each of the eight statements of the Accomplishment dimension represented their feelings about each gamification design. Besides the measurement of the perceived sense of accomplishment, the last question of the survey was “Which storyboard is your favorite?”. Thus, we were able to compare the perceived sense of accomplishment with the respondent’s preference for the gamification designs.

Before launching the survey, as recommended by Connelly (2008), we conducted a pilot study to assess whether the survey was being correctly understood by the respondents, as well as to assess whether the number of questions was adequate. This pilot group answered the survey before the survey application, with the question “Is this survey large?” added at the end of the survey. The pilot study was conducted with a small sample composed by 10 participants, where 80% answered that the survey wasn’t large so we decided to not take away any questions from it.
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3.2 Participants

The final survey was released on March 26, 2020, and was spread by social networks and e-mail. The survey was open for thirty-eight days and we received 366 answers, which 331 were valid according to our attention-check question. The respondents participated voluntarily, since we did not offer any kind of remuneration or gifts to the respondents. The study sample size is adequate under different aspects considering this type of study. According to the definitions of Bentler and Chou (1987) and Hair et al. (1998), it is necessary to have at least five participants for each construct measured (our study had seven constructs). Loehlin (1998) suggests a minimum sample of 100 participants for studies of this nature. Table 2 presents the demographic information of the respondents.

| Table 2 Demographic information |
|----------------------------------|
| Variable | % | Variable | % |
| Gender | | Age | |
| Female | 52 | 10–14 | 0.30 |
| Male | 47 | 15–19 | 9 |
| Other | 0.60 | 20–24 | 12 |
| Preferred not to answer | 0.60 | 25–29 | 15 |
| Education level | | |
| Elementary/Middle School | 2 | 35–39 | 13 |
| High School | 9 | 40–44 | 12 |
| Bachelor | 30 | 45–49 | 10 |
| Specialized courses | 21 | 50–54 | 7 |
| M.Sc. | 25 | 55–59 | 4 |
| Ph.D. | 8 | Over 60 | 1 |
| PostDoc | 4 | | |
| Gaming habits | | Frequency | |
| Play games | 67 | Every day | 13 |
| Do not play games | 33 | Every week | 21 |
| I do not know | 19 | Rarely | 47 |

Table 3 Participants distribution, average scores and standard deviation

| User types | D (%) | Mean score | S.D. | Female mean score | S.D. | Male mean score | S.D. |
|------------|-------|------------|------|-------------------|------|----------------|------|
| Philanthropist | 35 | 24.18 | 4.78 | 23.86 | 5.47 | 24.50 | 3.89 |
| Achiever | 30 | 23.98 | 4.79 | 23.41 | 5.60 | 24.57 | 3.61 |
| Free Spirit | 12 | 22.50 | 4.63 | 22.23 | 5.30 | 22.71 | 3.74 |
| Player | 12 | 20.53 | 5.61 | 19.76 | 5.96 | 21.37 | 5.05 |
| Socializer | 10 | 20.42 | 5.7 | 20.49 | 6.03 | 20.51 | 5.22 |
| Disruptor | 1 | 14.66 | 5.33 | 13.91 | 5.48 | 15.32 | 4.98 |

* D Distribution of the dominant user types, S.D. standard deviation

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Table 3 summarizes the participants’ distribution by the dominant user types (i.e., the strongest tendency of the participants), the average scores, and the standard deviation for each Hexad user type. Resembling the HEXAD results of Marczewski (2020), our research identified that Philanthropist is the most common dominant user type and Disruptor is the least common dominant user type. Comparing the female and male scores in each Hexad user orientations (i.e., the tendencies of the participants), it is possible to identify that the male scores were higher than female scores in all of the Hexad user orientations.

4 Results

To ensure the instrument validation for our study, we first analyzed the data normality (using the Shapiro–Wilk test as recommended by Wohlin et al. (2012)), which showed that our data followed a non-normal distribution. Then, we measured the internal reliability for each Hexad sub-scale (user types in the survey), as well as for the perceived sense of accomplishment evaluation in each storyboard. Overall, the reliability was acceptable (α ≥ 0.70, RHO A ≥ 0.70, CR ≥ 0.70, AVE ≥ 0.50) for all storyboards and user orientations, except for the Disruptors. We also measured the discriminant validity finding acceptable values, considering that the square root of the variables’ AVE value was larger than the correlations that the variable had with the other variables, and all of the variables presented correlations between them below 0.85. The reliability results are shown in Table 4, and the discriminant validity is shown in Table 5.

To answer our research question and measure the effects of the personalized gamification designs in terms of sense of accomplishment and preference, following
|       | AccE | AccF | AccPF | AccP | AccS | A    | D    | F    | P    | R    | PrE  | PrF  | PrPF | PrP  | PrS  | S    |
|-------|------|------|-------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|
| AccE  | 0.918|      |       |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| AccF  | 0.474| 0.889|       |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| AccPF | 0.614| 0.550| 0.919 |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| AccP  | 0.620| 0.623| 0.639 | 0.901|      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| AccS  | 0.652| 0.545| 0.638 | 0.574| 0.927|      |      |      |      |      |      |      |      |      |      |      |      |      |
| A     | 0.367| 0.400| 0.480 | 0.389| 0.429| 0.858|      |      |      |      |      |      |      |      |      |      |      |      |
| D     | 0.209| 0.246| 0.255 | 0.262| 0.254| 0.499| 0.653|      |      |      |      |      |      |      |      |      |      |      |
| F     | 0.268| 0.369| 0.369 | 0.294| 0.322| 0.740| 0.539| 0.760|      |      |      |      |      |      |      |      |      |      |
| P     | 0.353| 0.385| 0.441 | 0.385| 0.386| 0.771| 0.398| 0.696| 0.862|      |      |      |      |      |      |      |      |      |
| R     | 0.313| 0.280| 0.349 | 0.302| 0.369| 0.563| 0.425| 0.511| 0.413| 0.797|      |      |      |      |      |      |      |      |
| PrE   | -0.074| 0.142| -0.134| -0.094| -0.045| -0.042| -0.078| 0.015| 0.002| -0.041| 1.000|      |      |      |      |      |      |      |
| PrF   | -0.003| -0.079| 0.183| -0.070| -0.047| 0.045| 0.086| 0.048| 0.010| 0.034| -0.343| -0.260| 1.000|      |      |      |      |      |
| PrPF  | 0.030| 0.037| 0.030| 0.135| -0.014| -0.008| -0.044| 0.026| 0.025| 0.096| -0.153| -0.116| -0.248| 1.000|      |      |      |      |
| PrP   | -0.011| 0.077| -0.011| 0.053| 0.165| -0.008| 0.059| -0.061| 0.019| -0.074| -0.273| -0.207| -0.441| -0.197| 1.000|      |      |      |
| PrS   | 0.306| 0.444| 0.364| 0.356| 0.451| 0.559| 0.310| 0.541| 0.665| 0.393| -0.070| -0.024| -0.075| -0.048| 0.192| 0.858|      |      |

P: Philanthropist; A: Achiever; R: Player; F: Free Spirit; S: Socialzer; D: Disruptor; AccE: Perceived sense of accomplishment for the storyboard Ecological; AccF: Perceived sense of accomplishment for the storyboard Fictional; AccPF: Perceived sense of accomplishment for the storyboard Performance; AccP: Perceived sense of accomplishment for the storyboard Personal; AccS: Perceived sense of accomplishment for the storyboard Social; PrE: Preference Ecological; PrF: Preference Fictional; PrPF: Preference Performance; PrP: Preference Personal; PrS: Preference Social
other recent studies in personalized gamification (Orji et al. 2018; Hallifax et al. 2019b; Stuart et al. 2020), we employed the Partial Least Squares Path Modeling (PLS-PM) analysis to identify the relation between the Hexad user types with the gamification designs (sense of accomplishment and preference), since it is a reliable method for estimate cause-effect relationship models with latent variables (Hair Jr et al. 2016). To perform the statistical analysis in our study, we used SPSS 26 software. To conduct the PLS-PM, we used SmartPLS3 software, that provides a graphical interface to calculate PLS-PM (Wong 2013). Our complete dataset can also be found in the complementary files.

Table 6  Comparison between the favorite storyboard and the perceived sense of accomplishment

| Storyboard   | Preference (%) | Accomplishment (%) | Female Acc (%) | Male Acc (%) |
|--------------|---------------|--------------------|----------------|--------------|
| Fictional    | 11            | 14                 | 15             | 13           |
| Personal     | 10            | 12                 | 13             | 11           |
| Performance  | 36            | 28                 | 27             | 30           |
| Ecological   | 18            | 21                 | 20             | 22           |
| Social       | 26            | 25                 | 25             | 25           |

Preference: Which storyboard did you prefer?; Accomplishment: Perceived sense of accomplishment; Female Acc: Perceived sense of accomplishment by the female respondents; Male Acc: Perceived sense of accomplishment by the male respondents

Fig. 2  Research model

https://www.smartpls.com/.
Table 6 presents the preference and perceived sense of accomplishment averages in general and by gender. It is possible to identify that the Performance gamification design is the most chosen in terms of preference and perceived sense of accomplishment.

The research model of our study is shown in Fig. 2. The results indicate that the Achiever orientation is positively associated with perceived sense of accomplishment from the Performance ($\beta = 0.295^{***}$), and Social ($\beta = 0.238^*$) designs. The Player orientation is positively associated with perceived sense of accomplishment from Ecological ($\beta = 0.162^*$) and Social ($\beta = 0.161^*$) designs; positively associated with preference from Personal ($\beta = 0.165^*$) design; and negatively associated with preference from Social ($\beta = -0.148^*$) design. The Free Spirit orientation is only negatively associated with preference from Social ($\beta = 0.150^*$) design. The Socializer orientation is positively associated with perceived sense of accomplishment from Fictional ($\beta = 0.307^{***}$), Personal ($\beta = 0.154^*$) and Social ($\beta = 0.309^{***}$) designs; positively associated with preference from Social ($\beta = 0.370^{***}$) design; and negative associated with preference from Performance ($\beta = -0.165^*$) design. The Disruptor orientation is positively associated with preference from Social ($\beta = 0.150^*$) design. The Philanthropists orientation did not present any association. All of the relations are shown in Table 7.

4.1 Discussion

The goal of this study was to understand the relationship between user orientations (Achiever, Disruptor, Free Spirit, Philanthropist, Player, and Socializer) and gamification designs (Fictional, Personal, Performance, Ecological, and Social). Overall, we identified different positive and negative associations between five of the six user orientations with the gamification designs. However, we identified that there is no consistent pattern of associations.

In terms of the user type distribution, our results (see Table 3) are similar to the results of Marczewski (2020), with Philanthropist as the most common dominant user type and Disruptor as the least common dominant user type. Also, our results are similar to the results found by Tondello et al. (2019), where they identified that Philanthropist and Achiever are the prevalent user orientations and Disruptor was the one that scored lower. According to Tondello et al. (2019), women tend to score higher than men in all the user orientations with intrinsic motivations (i.e., Philanthropist, Socializer, Free Spirit, and Achiever), and even though in our results men scored higher than women in all the user orientations, the difference was smaller in the Philanthropist, Socializer, Free Spirit, and Achiever orientations (i.e., the user orientations that are motivated intrinsically).

Starting to answer our research question, considering the effects of personalized gamification designs on user orientations’ sense of accomplishment and preference (see Table 7), similar with the results found by Hallifax et al. (2019b) and Tondello et al. (2016), our results showed that Philanthropists did not present a significant association with any gamification design. Moreover, our results showed that Philanthropists presented a negative association with the Fictional gamification design in
Table 7  Results: associations between gamification designs and user orientation

|                | β     | P-value | CI          |                | β     | P-value | CI          |
|----------------|-------|---------|-------------|----------------|-------|---------|-------------|
|                |       |         | 2.5% | 97.5% |       |         | 2.5% | 97.5% |
| PAcc → SF      | −0.006 | 0.951 | −0.176 | 0.198 | AAcc → SF | 0.157 | 0.110 | −0.035 | 0.333 |
| PPr → SF      | −0.029 | 0.765 | −0.214 | 0.172 | APr → SF | 0.080 | 0.444 | −0.121 | 0.281 |
| PAcc → SP      | 0.164 | 0.063 | −0.029 | 0.342 | AAcc → SP | 0.168 | 0.066 | −0.027 | 0.314 |
| PPr → SP      | −0.013 | 0.890 | −0.209 | 0.171 | APr → SP | −0.064 | 0.515 | −0.234 | 0.105 |
| PAcc → SPF     | 0.149 | 0.082 | −0.028 | 0.304 | AAcc → SPF | 0.295*** | 0.001 | 0.119 | 0.456 |
| PPr → SPF     | 0.020 | 0.852 | −0.193 | 0.216 | APr → SPF | 0.047 | 0.667 | −0.159 | 0.255 |
| PAcc → SE      | 0.166 | 0.069 | −0.004 | 0.338 | AAcc → SE | 0.174 | 0.075 | −0.012 | 0.346 |
| PPr → SE       | 0.119 | 0.242 | −0.086 | 0.317 | APr → SE | −0.113 | 0.315 | −0.329 | 0.079 |
| PAcc → SS      | 0.005 | 0.956 | −0.146 | 0.176 | AAcc → SS | 0.238* | 0.011 | 0.049 | 0.420 |
| PPr → SS       | −0.095 | 0.309 | −0.261 | 0.056 | APr → SS | 0.034 | 0.724 | −0.143 | 0.214 |
| RAcc → SF      | 0.030 | 0.627 | −0.095 | 0.154 | FAcc → SF | 0.058 | 0.429 | −0.082 | 0.206 |
| RPr → SF      | 0.042 | 0.613 | −0.115 | 0.182 | FPr → SF | −0.021 | 0.804 | −0.210 | 0.133 |
| RAcc → SP      | 0.199 | 0.101 | −0.031 | 0.241 | FAcc → SP | −0.132 | 0.127 | −0.291 | 0.062 |
| RPr → SP       | 0.165** | 0.002 | 0.048 | 0.261 | FPr → SP | 0.104 | 0.261 | −0.093 | 0.298 |
| RAcc → SPF     | 0.117 | 0.062 | −0.004 | 0.230 | FAcc → SPF | −0.063 | 0.351 | −0.193 | 0.081 |
| RPr → SPF      | 0.009 | 0.892 | −0.124 | 0.123 | FPr → SPF | 0.040 | 0.661 | −0.139 | 0.244 |
| RAcc → SE      | 0.162* | 0.014 | 0.033 | 0.310 | FAcc → SE | −0.124 | 0.142 | −0.281 | 0.036 |
| RPr → SE       | −0.005 | 0.946 | −0.138 | 0.140 | FPr → SE | 0.145 | 0.066 | −0.032 | 0.276 |
| RAcc → SS      | 0.161* | 0.012 | 0.021 | 0.291 | FAcc → SS | −0.128 | 0.154 | −0.299 | 0.037 |
| RPr → SS       | −0.148* | 0.031 | −0.270 | −0.008 | FPr → SS | −0.226** | 0.009 | −0.386 | −0.073 |
| SACc → SF      | 0.307*** | 0.000 | 0.171 | 0.437 | DACc → SF | 0.031 | 0.644 | −0.091 | 0.154 |
| SACc → SP      | −0.021 | 0.803 | −0.190 | 0.127 | DACc → SP | −0.108 | 0.189 | −0.260 | 0.036 |
| SACc → SPF     | 0.154* | 0.036 | 0.004 | 0.297 | DACc → SPF | 0.090 | 0.231 | −0.055 | 0.239 |
| SACc → SE      | −0.092 | 0.284 | −0.259 | 0.083 | DACc → SE | −0.104 | 0.166 | −0.250 | 0.030 |
| SACc → SPF     | 0.087 | 0.204 | −0.042 | 0.231 | DACc → SPF | 0.006 | 0.916 | −0.085 | 0.123 |
| SACc → SE      | −0.165* | 0.032 | −0.305 | −0.025 | DACc → SE | 0.081 | 0.258 | −0.072 | 0.223 |
| SACc → SF      | 0.094 | 0.186 | −0.052 | 0.229 | DACc → SF | 0.026 | 0.732 | −0.115 | 0.174 |
| SACc → SE      | −0.130 | 0.081 | −0.270 | 0.005 | DACc → SE | −0.104 | 0.139 | −0.222 | 0.025 |
The relationship between user types and gamification designs

Terms of perceived sense of accomplishment and preference, the only design that did not present the “assistant” that explained what the student would do in that gamification design. Since the Philanthropists are motivated by interaction with others (Tondello et al. 2016), we believe the lack of the “assistant” presence can be understood by Philanthropists as a lack of interaction.

Analyzing our results in comparison with other studies (Tondello et al. 2016, 2017a), we believe that Achievers had a strong significant association with the Performance and Social gamification designs in the perceived sense of accomplishment measurement, especially because these two designs showed game elements and situations that could lead the user to feeling achievement and to demonstrate competence, which intrinsically motivates this user type (Tondello et al. 2016). We believe that when implemented in a gamified system, the Performance (game elements: Level, Point, Progression, Stats, and Acknowledgement) and Social (game elements: Social Pressure, Competition, Social Status, and Cooperation) gamification designs would probably lead the Achievers to have a feeling of advancement in their skills, and thus motivate them.

Since Players are motivated by extrinsic rewards (Tondello et al. 2016), we believe that the significant associations with the Ecological gamification design were related to the game elements of Rarity and Economy. Orji et al. (2018) found that Players tend to be motivated by Competition and Cooperation, which can explain the positive significant association with the Social gamification design in terms of perceived sense of accomplishment. At the same time, Players presented a slight and negative significant association with the Social gamification design in terms of preference. Thus, even though the Social gamification design brought a sense of accomplishment to the Players, this gamification design was not preferred by them. Players also presented a positive significant association with the Personal gamification design, probably because the game element Puzzle (Challenge), related to this user type in other studies (Tondello et al. 2017a, 2016).

Free Spirits only presented one significant association with the Social gamification design; however, it was negative. Also, this user orientation was the one that presented more negative associations, since we were able to identify that Free

Table 7 (continued)

| β  | P-value | CI 2.5% | CI 97.5% | β  | P-value | CI 2.5% | CI 97.5% |
|----|---------|---------|----------|----|---------|---------|----------|
| SAcc → SS | 0.309*** 0.000 | 0.155 | 0.446 | DAcc → SS | 0.038 | 0.564 | − 0.077 | 0.165 |
| SPr → SS | 0.370*** 0.000 | 0.259 | 0.467 | DPr → SS | 0.150* | 0.017 | 0.023 | 0.254 |

*p < 0.1, **p < 0.05, ***p < 0.01; β: Regression Coefficient; CI: Confidence Interval; PAcc: Philanthropist perceived sense of Accomplishment; PPr: Philanthropist Preference; AAcc: Achiever perceived sense of Accomplishment; APr: Achiever Preference; RAcc: Player perceived sense of Accomplishment; RPr: Player Preference; FAcc: Free Spirit perceived sense of Accomplishment; FPr: Free Spirit Preference; SAcc: Socialzer perceived sense of Accomplishment; SPr: Socialzer Preference; DAcc: Disruptor perceived sense of Accomplishment; DPr: Disruptor Preference; SF: Storyboard Fictional; SP: Storyboard Personal; SPF: Storyboard Performance; SE: Storyboard Ecological; SS: Storyboard Social
Spirits presented negative non-significant associations with all of the gamification designs. Considering preference, this user orientation presented negative non-significant associations with Fictional gamification design, and considering perceived sense of accomplishment, negative non-significant associations with Personal, Performance, Ecological, and Social gamification designs. This was unexpected considering that we presented game elements in this study that were related to this user type in previous studies (e.g. Puzzle (Tondello et al. 2017a, 2016) and Level (Tondello et al. 2017a)).

Socializers were the user orientation that presented more significant associations, including a strong significant association with the Social gamification design. The game elements presented in the Social gamification design are important to ensure interactions between the users (Toda et al. 2019) and can be related directly with the Socializers that are intrinsically motivated by relatedness (Tondello et al. 2019). We understand that the Social gamification design, when implemented in a gamified system, could guide the Socializers into relatedness, which would motivate them. The game elements of the Social gamification design (game elements: Social Pressure, Competition, Social Status, and Cooperation) have already been individually associated with Socializers in previous studies (Tondello et al. 2016; Marczewski 2017; Tondello et al. 2017a). Probably, the strong significant association with the Fictional gamification design occurred because the game element Narrative is related to the user’s interaction with the system (Toda et al. 2019), and the slight significant association with the Personal gamification design because of the game element Puzzle (Challenge), that has been related with this user orientation before (Tondello et al. 2016). They also presented a slight and negative significant association with the Performance gamification design, probably because of the game elements showing progress in this gamification design, considering that similar results were found by Hallifax et al. (2019b).

We believe Disruptors presented a significant association with the Social gamification design, especially because of the game element Competition. Tondello et al. (2016) identified that competition is a game element that can be related with this user orientation, and Orji et al. (2018) showed that competition would motivate people with high disruptor tendencies. Also, the Social gamification design is the one that shows interactions with other students, and Disruptors need interactions to influence other users to try to change the system (Marczewski 2015). Hence, this can be another reason for this significant association.

Our results also presented some non-significant associations between the user orientations and the gamification designs that indicate some possibilities. As aforementioned, the Philanthropists presented a non-significant association with the Fictional gamification design. We believe that designers should guarantee that, especially when the gamified system does not present interactions with other users, they have at least a figure to simulate interaction (e.g., animated pedagogical agents or an “assistant” from the system). Considering the non-significant associations, Players presented a positive association in terms of perceived sense of accomplishment with Performance and Personal gamification designs. Since this user orientation also presented a significant association with the Personal gamification design in terms of preference, we understand that the implementation of this design would be a good
option to increase the motivation of this user type. The association with the Performance gamification design could be explained by the use of the game elements Level and Point, related to this user type before (Tondello et al. 2016, 2017a). Therefore, the implementation of this design to Players also could be a good option to increase the sense of accomplishment in these users. The Disruptors presented a negative association with the Ecological, Personal, and Fictional gamification designs. The Personal and Fictional designs represented how the system would work; therefore, this could be seen by the Disruptors as the boundaries of the system. The game element Time Pressure, represented by a clock in the Ecological design, could also be seen by the Disruptors as a limitation of their actions. Since they are motivated by change (Tondello et al. 2016), we understand these designs could be understood by Disruptors as limiting, which could explain these negative associations.

Our results show that the perceived sense of accomplishment measurement has more homogeneous results than the preference measurement (see Table 6). This demonstrates that only measuring the preference for game elements might not be sufficient to understand the effects of the game elements in user experience. For instance, some user orientations presented a significant association with a gamification design in terms of perceived sense of accomplishment, but this not has happened in terms of preference with the same gamification design. Considering that the feeling of accomplishment drives the user to complete tasks or goals and reflects the user’s engagement (Högberg et al. 2019), we believe the user orientations that presented a significant association with a gamification design in terms of perceived sense of accomplishment can present better progress when using gamified systems that have that set of game elements.

When we do not consider the user orientation, the game elements of the Performance and Social gamification designs can be used for all users, since these two designs showed a predominance in the preference and perceived sense of accomplishment results (see Table 6). Thus, gamified systems must present the group of game elements that create interactions between the users, and the group of game elements that provide feedback to them.

Our results corroborate other studies (Mora et al. 2019; Hallifax et al. 2019b; Tondello et al. 2017a) showing that users have different preferences based on their user orientations. Considering only preference by user orientation, the Social gamification design is strongly related with Socializers and can also be used with Disruptors, while the Personal gamification design can be indicated for Players. Furthermore, the user orientation is a factor that affects how the users perceived their sense of accomplishment. Considering the perceived sense of accomplishment and the user orientations, the Social and Performance gamification designs are the most related to Achievers, and Socializers seems to be strongly affected by the Social and Fictional gamification designs. Philanthropists seem not to be affected by the game elements represented in this study, and Free Spirits might not be positively affected by most of the game elements.

According to our results, it is possible to select the most appropriate gamification designs for each system user, according to their Hexad user type and based on two different approaches (preference and/or perceived sense of accomplishment). This can help designers to personalize gamification, and therefore positively affect
the users. To personalize gamified environments for people with higher Achiever tendencies, designers should focus on implementing the game elements from the Performance and Social gamification designs, since our results indicated a significant association in terms of perceived sense of accomplishment. The game elements from the Ecological and Personal gamification designs should be avoided since they presented a negative non-significant association. For people with higher Disruptor tendencies, designers should focus on implementing the game elements from the Social gamification design, especially the game element Competition. Considering the non-significant associations this user orientation presented, if the gamified environment is based on the preference for game elements, it is important to avoid or use with caution the game elements from the Ecological, Personal, and Fictional gamification designs.

For people with higher Player tendencies, designers can focus on the game elements from the Personal and Ecological gamification designs. Since Players presented negative (preference) and positive (sense of accomplishment) significant associations with the Social gamification design, designers can give to these users the possibility to choose if they want to interact with the game elements from this gamification design. For people with higher Socializer tendencies, designers should focus on implementing the game elements from the Social gamification design, since this user type presented a significant association with this gamification design in both approaches (preference and sense of accomplishment). The game elements from the Personal and Fictional gamification designs can also be implemented to increase the experience for this user type. Game elements from the Performance gamification design should be avoided.

Since Free Spirits only presented a negative significant association with the Social gamification design, we indicate that the game elements from this gamification design should be avoided or used with caution. Considering the negative and non-significant associations this user type presented with the other gamification designs, we understand that designers could give users with high Free Spirit tendencies, a possibility to disable the game elements or to freely choose which game element each user would interact with. The Philanthropists did not present any significant association, therefore, designers can use the non-significant associations

| Table 8 | Recommendations to personalize gamification |
|---------|---------------------------------------------|
| Preferece | Sense of accomplishment |
| Philanthropist | Ø | Ø |
| Achiever | Ø | + SPF and + SS |
| Player | − SS and + SP | + SE and + SS |
| Free Spirit | − SS | Ø |
| Socializer | − SPF and + SS | + SF, + SP and + SS |
| Disruptor | + SS | Ø |

Ø: Without significant association; +: Significant positive association; −: Significant negative association; SF: Storyboard Fictional; SP: Storyboard Personal; SPF: Storyboard Performance; SE: Storyboard Ecological; SS: Storyboard Social
presented in our results to personalize gamified environments to this user type or also implement the possibility to disable the game elements or to freely choose which game element each user would interact with. Furthermore, we indicate that the system should provide interaction with other users or at least with the system itself, through an “assistant”. In Table 8, we summarize these recommendations of which gamification designs can be used to personalize gamified systems based on the significant associations we found (see Table 7).

4.2 Limitations

Some limitations have emerged during the study and they need to be considered. Although the internal reliability for the Disruptors was below the acceptable threshold, we were able to identify that for this user orientation, there exists a kind of predominance: all of the Disruptors, except one, presented only this user orientation as the dominant user type, and we were not able to find any study that has reported similar results. As another observation, the use of gamification designs can bring different results from the use of a real gamified system. Our design focused on representing different phases a gamified system would have, thus, they were a design concept of a gamified system. While storyboards represent the “ideal” scenario to evaluate a design idea, the implementation brings other design decisions (Orji et al. 2014), which could influence the users’ response. Also, two of the storyboards did not represent the moment the student would answer questions in the system, considering they were designed to show how a student would create a profile and also the moment that the student would know the system. Some respondents could see this as a gap, which could subsequently influence their responses in the survey.

We have sought to mitigate some of the foreseeable limitations during the conduct of the study. To mitigate the possibility that the storyboards do not represent the dimensions proposed by Toda et al. (2019), we validated the storyboards with three gamification experts before using them in the survey. The size of the survey could have led people to answer without paying attention, and to mitigate this threat, we used an “attention-check” question in the survey and eliminated the responses that did not pass this validation. As a final action, since the Hexad scale proposed by Tondello et al. (2019) and the Gameful Experience Questionnaire proposed by Höberg et al. (2019) were not empirically validated in Brazilian Portuguese, we conducted a statistical analysis to validate the answers obtained in our study to mitigate this threat.

4.3 Recommendations for future studies

Based on the results obtained in our study, as well as the limitations our study raised, it is possible to propose a series of new studies to deepen this research domain. Initially, our study focused on answering research questions in the field of education (i.e., using storyboards representing a gamified setting). At the same time, the effects of gamification may vary according to the field of application (e.g., marketing, health, addictions, and others). Thus, we believe that future
studies should be conducted in different areas (i.e., replicating our study in different domains) expanding our results by using specific gamification designs suited to the context.

Almost all the respondents were older than 15 years, which can prevent the results’ generalization for younger people. We believe that future studies should focus on people under 15 years old, analyzing if the age can change the preference and perceived sense of accomplishment, thus expanding our results. Since most of the respondents had reported that they play games (67%), an interesting perspective future studies should focus on is whether the gaming habit affects the preference and perceived sense of accomplishment in gamified systems.

In our study, we chose to conduct an exploratory study which allowed us to have a broad view of the subject, without exercising control over our subjects. However, now that our results provide us with an overview of the subject, further studies must deepen the results through experimental studies in controlled environments, for example, directly comparing two gamification designs (personalized vs. non-personalized) in terms of user experiences. Therefore, we recommend that future studies should conduct experiments comparing the effects of personalized with non-personalized gamification designs in the user types.

In our study, we measured one dimension of the gameful experience (i.e., Accomplishment), and according to Högberg et al. (2019), the dimension of Immersion also seems to reflect user’s engagement. Thus, future studies can be carried out to measure this dimension. This decision to look at the Accomplishment dimension of the gameful experience was important to obtain a reliable result for a specific dimension of the gameful experience. Now that we have these results, it may be necessary to assess the effects on other gameful experience dimensions (e.g., Challenge, Competition, and Playfulness). Thus, we recommend that future studies should investigate other gameful experience dimensions.

Our results have allowed us to identify different significant (positive and negative) relationships between different user orientations and gamification designs. This means that future studies on the personalization of gamification can use these results as a basis for personalizing gamified environments. At the same time, it is also important to investigate whether the results obtained in this study are maintained in ecological environments (i.e., real gamified systems). Thus, we recommend that future studies can implement the gamification designs proposed and validated in our study in gamified systems, and evaluate the effects of different versions of the system on the users’ experience during the system usage.

Also based on the results obtained in our study, it is possible to understand which gamification designs are more or less effective concerning the users’ sense of accomplishment. This can be useful, for example, to enable gamified system designers to personalize the gamification in a way that positively affects users. To make this task easier for designers, we recommend that future research may propose recommender systems (Resnick and Varian 1997) to suggest the most suitable gamification design for each user according to their Hexad user type.
5 Concluding remarks

In this study, we presented how the six Hexad user orientations are related with five different gamification designs. To avoid the use of only the game elements that are considered the most used in research (e.g., points and badges), the game elements used in these designs were chosen from an empirically validated gamification taxonomy that groups twenty-one game elements in five dimensions. We used validated storyboards to show these dimensions to the respondents, in order to help people visualize how each dimension could be represented in a gamified system, instead of only asking about the preference for a particular dimension. Furthermore, we compared the preference with the perceived sense of accomplishment (a dimension of the gameful experience). Our results corroborate other research in identifying that the game elements presented in the Performance (Point, Progression, Level, Stats, and Acknowledgment) gamification design can be considered most adequate for all users, and that the Hexad user orientations have different preferences concerning gamification designs. Also, our results showed that the game elements presented in the Fictional (Narrative and Storytelling) and Personal (Objective, Puzzle, Novelty, Sensation and Renovation) gamification designs were the least preferred by the respondents. Our findings showed how some gamification designs can be used according to the user orientation, helping designers to design personalized gamified systems. For future studies, we intend to focus on the measurement of the other dimensions of the gameful experience, and how users that present more than one dominant user type are affected by the gamification designs.

Appendix

See Table 9 and Figs. 3, 4, 5, 6 and 7.
Table 9  Storyboards description

| Dimension          | Description                                                                                                                                 |
|--------------------|---------------------------------------------------------------------------------------------------------------------------------------------|
| Fictional (SF)     | This gamification design represents what would be the initial page of a gamified educational system. The student would hear an audio message explaining that a teacher spread questions along a path in a forest, so as not to get lost (**game element Storytelling**). The student would be able to create a profile in the system, where they could create an avatar and share it in social networks, or start answering the questions without access to any bonus (**game element Narrative**) |
| Personal (SP)      | This gamification design represents what would be the objectives and configurations of a gamified educational system. The student’s goal would be to correctly answer twenty-one questions about certain content. This goal would be shown on the first screen of the gamified educational system (**game element objective**). When missing a question about a certain subject, the student would have the option to do another question about the same subject and earn the lost point of the wrong question (**game element renovation**). After three correct questions, the student would have the chance to answer a puzzle that would give them an extra life in the phase (**game element Puzzle**). Every time the environment would have an update (**game element novelty**), the user would receive an audible, visual, or vibration notification when logging into the system, according to the configuration he/she chooses (**game element sensation**) |
| Performance (SPF)  | This gamification design represents how a gamified educational system would respond to the student’s actions. The student would have to answer seven questions correctly to level up in the system and be a Beginner, Apprentice, or Master (**game element Level**). When the student answers the questions correctly, he/she would receive XP’s (**game element Point**) and would advance in the progression bar, represented by stars (**game element Progression**). When the student answers more than ten questions correctly, he/she would receive a recognition trophy (**game element Acknowledgement**). All of this information would be available on the page called “My progress” in the gamified educational system (**game element Stats**) |
| Ecological (SE)    | This gamification design represents how a gamified educational system would engage the student to follow a desired behavior. The student would have to choose a specific path to follow (**game element Imposed Choice**). After choosing the path, the student would spin a “wheel of luck” to earn a bonus, which could be “Skip a question”, “Request a help letter” or “Increase the level of the phase by one hour” (**game element Chance**). From there, they would have a limited time to finish the phase (**game element Pressure Time**), where if they finish in the proposed time, they would earn a rare stone (**game element Rarity**). The student would have the opportunity to exchange the item for more lives or items (**game element Economy**) |
| Social (SS)        | This gamification design represents how a gamified educational system would provide social interaction. The student would have to join other students on a group mission, where they could help each other in order for everybody to reach the end (**game element Cooperation**). The team which finishes first (**game element Competition**) would win the title “Pioneers” (**game element Reputation**). Participants would be notified whenever other teams were close to reaching them (**game element Social pressure**) |
Frame 1: Imagine you’re entering a system for the first time and hear an audio that says...; Frame 2: “A teacher was walking in a forest and filled the way with challenges, so as not to get lost on the way back. Are you able to unravel them to help her get home? You can create a profile or reply anonymously. What do you prefer?”; Frame 3: “If you chose to continue without creating a profile in the system, you can now start answering the questions”; Frame 4: “If you chose to create a profile, choose one avatar below.”; Frame 5: “This will be your avatar in the system. You can post it on your social networks.”; Frame 6: “Now you can start answering the questions. Good luck!”

Fig. 3 Storyboard fictional
Frame 1: Hello, welcome to our system! How about taking a tour to know the rules and icons of the system?

Frame 2: Your goal is to answer 21 questions correctly.

Frame 3: Every three correct questions you have the chance to answer a puzzle that will give you an extra life!

Frame 4: Don’t worry! If you miss a question, you’ll have the option to answer another question about the same topic and earn the lost point.

Frame 5: Every time the system has an update, you’ll receive a notification when you log into the system. It can be auditory, visual, or vibration, according to your preferences. This way you won’t miss any news! You can also propose improvements in the system.

Frame 6: Our tour is over. Now you can start answering the questions. Good luck!

Fig. 4 Storyboard personal
Frame 1: Welcome to the system! On the way, you'll find boxes with questions to be answered. You'll reach the “Beginner” level when you get seven questions right. Good luck!；Frame 2: You've answered the first question correctly and won 10 XPs!；Frame 3: -；Frame 4: Congrats, you've reached the Beginner level! When you answer fourteen questions correctly, you'll reach the Apprentice level.；Frame 5: You have answered ten questions correctly and won your first trophy!；Frame 6: Welcome to your progress page! Here you can see what you have already won and what you can still win in the system.

Fig. 5 Storyboard performance
Frame 1: Hello! This phase is a group challenge. Júlia and John are on this mission with you. You need to answer five questions correctly to finish the mission.; Frame 2: Attention! John is in trouble and if he doesn’t finish question three, the group won’t finish the mission. Do you want to send a help letter to John?; Frame 3: John answered question three correctly! Just two more questions and you’ll finish the mission!; Frame 4: Group, attention! There are two other groups finishing the fourth question and they can surpass you in the challenge.; Frame 5: Congratulations team! You are the first team to finish the challenge and have won first place in the class!; Frame 6: This is the title that the group received for finishing in first place.

Fig. 6 Storyboard social
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Author Contributions ACGS took part in conceptualization, methodology, investigation, data curation, formal analysis, writing—original draft. WO involved in conceptualization, methodology, supervision, formal analysis, validation, writing—review & editing. JH participated in supervision, validation, formal analysis, writing—review & editing. LR involved in writing—review. PTP took part in writing—review. AMT participated in writing—review. SI involved in supervision, writing—review.

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Data availability Original dataset available as supplementary material.

Fig. 7 Storyboard ecological

Frame 1: To finish this phase you need to choose which path you will follow.; Frame 2: Attention! You chose the left path and now you can spin the wheel of luck to get a bonus!; Frame 3: Congratulations! You can skip a question from this phase.; Frame 4: Attention! Now you have an hour to finish this phase. Good luck!; Frame 5: You have finished the phase in less than an hour and got a rare stone! It can be exchanged for other items.; Frame 6: - .

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Ana Cláudia Guimarães Santos has a B.Sc. in information systems, an MBA in software engineering, and is a specialist in computers applied to education from the University of São Paulo (USP). Her main research interests are gamification, educational policy, educational technologies, and software engineering. Apart from research, since 2014 she is a Brazilian government employee, creating, coordinating, and ministering lectures, training, and courses to teachers from public schools on how to use technology in class.

Wilk Oliveira is an assistant professor in the University of São Paulo (Brazil) and was a visiting lecturer in the Tiradentes University Center (Brazil). He is a researcher at the Laboratory of Computing Applied to Education and Advanced Social Technology (University of São Paulo), Gamification Group (Tampere University—Finland), and the Learning Lab (Durham University—UK). Wilk collaborated with the Brazilian Ministry of Education working on projects related to the design, application, and evaluation of educational technologies. Wilk Oliveira has been working on a series of research projects, maintained by important international funding agencies, where he has worked on the design, application, and evaluation of various educational technologies, generating different products and publishing numerous scientific studies in some of the main international conferences and journals related to Educational Technologies. Even as a young researcher, Wilk has already received several awards, including best paper award at different international conferences. His main research interests are educational technology (especially, but not exclusively: flow theory, gamification, and educational games) and computer science education.

Juho Hamari is a professor of gamification and the leader of the Gamification Group at the Faculty of Information Technology and Communication Sciences at Tampere University. Professor Hamari is leading research on gamification under the Academy of Finland Flagship, Profiling, and Center of Excellence programs. His and his group’s research covers several forms of information technologies such as games, motivational information systems, new media (social networking services, eSports), peer-to-peer economies (sharing economy, crowdsourcing), and virtual economies. Dr. Hamari has authored several seminal empirical, theoretical, and meta-analytical scholarly articles on these topics from perspective of consumer behavior, human–computer interaction, game studies, and information systems science.

Luiz Rodrigues is a researcher at the Laboratory of Applied Computing to Education and Advanced Social Technology (CAEd) and a part-time lecturer at the Faculty of Industries from Londrina, Brazil. Has a B.Sc. and M.Sc., and is a Ph.D. candidate at the University of São Paulo, all in Computer Science. His research interests are gamification, personalization and user modeling, educational data mining, educational games, and procedural content generation.

Armando M. Toda is a PDRA at Durham University and research collaborator at University of São Paulo (USP). Has a B.Sc., M.Sc. and Ph.D. in computer science, the latest from the University of São Paulo. His research career is invested in the design and application of gamification in learning environments. His scientific and social mission is to demonstrate how game-based technologies can improve people’s life and how these technologies can impact positively in education. Has published many scientific papers in journals and conferences, which won international recognition and awards. His research interests are aligned but not limited to: gamification, human–computer interaction, learning environments, artificial intelligence applied to education, educational data mining, computational thinking.

Paula T. Palomino is a Ph.D. Candidate in Computer Science at University of São Paulo (USP) and an international visiting graduate student at HCI Games Group/the University of Waterloo. She holds B.Sc. and M.Sc. in Communication from Federal University of São Carlos (UFSCar). Her main research areas are game design, gamification, UX and digital culture, specifically working with the use of subjective game elements in gamification frameworks applied to educational contexts.

Seiji Isotani is a full Professor of computer science and learning technology at the Institute of Mathematics and Computer Science of the University of Sao Paulo, Brazil. He received B.Sc. and M.Sc. degrees
in computer science from the University of Sao Paulo, Brazil, and a Ph.D. degree in information engineering from Osaka University, Japan. His research career has been devoted to the conception, design, development, testing and deployment of intelligent, and collaborative educational systems using ontologies and other semantic technologies. His scientific and social missions converge into a single objective: to enable the realization of Anytime, Anywhere, Anybody Learning (AAAL) by developing cutting-edge technology. He is the co-founder of two startups, which have won several innovation awards in the fields of education and the semantic Web. He has published over a hundred scientific papers, books, and book chapters on educational technology and the semantic Web. His mains research interests are in the areas of gamification, ontological engineering, computer-supported collaborative learning (CSCL), artificial intelligence in education (AIED), and technology-enhanced learning.