



An Immersive Self-Report Tool for the Affective Appraisal of 360° VR Videos

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Immersive 360° virtual reality (VR) movies can effectively evoke a wide range of different emotional experiences. To this end, they are increasingly deployed in entertainment, marketing and research. Because emotions influence decisions and behavior, it is important to assess the user's affective appraisal of immersive 360° VR movies. Knowledge of this appraisal can serve to tune media content to achieve the desired emotional responses for a given purpose. To measure the affective appraisal of immersive VR movies, efficient immersive and validated instruments are required that minimally interfere with the VR experience itself. Here we investigated the convergent validity of a new efficient and intuitive graphical (emoji-based) affective self-report tool (the EmojiGrid) for the assessment of valence and arousal induced by videos representing 360° VEs (virtual environments). Thereto, 40 participants rated their emotional response (valence and arousal) to 62 videos from a validated public database of 360° VR movies using an EmojiGrid that was embedded in the VE, while we simultaneously assessed their autonomic physiological arousal through electrodermal activity. The mean affective ratings obtained with the EmojiGrid and those provided with the database (measured with an alternative and validated instrument) show excellent agreement for valence and good agreement for arousal. The mean arousal ratings obtained with the EmojiGrid also correlate strongly with autonomic physiological arousal. Thus, the EmojiGrid appears to be a valid and immersive affective self-report tool for measuring VE-induced emotions.

Keywords: 360° VR, immersive VR, valence, arousal, emotions, EmojiGrid

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INTRODUCTION

Motivation of This Study

Interactive and immersive VR movies and virtual environments can effectively evoke a wide range of different emotions (Felnhofer et al., 2015; Ding et al., 2018; Chirico and Gaggioli, 2019). As a result, these media are increasingly deployed in entertainment, marketing and research. However, despite their increasing popularity, it is still not clear how the VE content should be designed to achieve the desired emotional responses of users (Riva et al., 2007). To explore this relation, we need efficient and validated affective self-report instruments to measure the affective responses to VR experiences (Oliveira et al., 2018). In this study we investigate the convergent validity of a new affective self-report tool for the assessment of immersive VE-induced emotions.

According to the circumplex model of affect (Russell, 1980), emotions can be described by their two main principal dimensions: valence (pleasantness) and arousal. Valence refers to the degree of positive or negative affective response to a stimulus, while arousal refers to the intensity of the

affective response (i.e., the degree of activation or deactivation). This suggests that participants can report their emotional state by responding along these two dimensions.

Various methods are currently available to assess the affective quality of VR experiences. These methods can be classified as objective and subjective measures (for an overview and discussion see Skarbez et al., 2017). *Objective measures* include physiological measures (e.g., heart rate variability, pupil size variability, electrodermal activity, electrocardiogram data, electromyogram data, electroencephalogram data, functional magnetic resonance data) and behavioral measures (e.g., gaze behavior, reflexive responses, postural sway). Physiological measures are objective, but also more elaborate and typically require extensive post-processing and interpretation (Brouwer et al., 2015). *Subjective measures* are typically obtained through questionnaires and rating tools. Although these instruments are typically intrusive, they are still the preferred method of investigation since they are cheap and easy to administer in almost any condition (Grassini and Laumann, 2020).

We recently introduced the EmojiGrid (Toet et al., 2018; see section The EmojiGrid; Kaneko et al., 2018b): an emoji-based graphical self-report tool to assess valence and arousal. The tool is intuitive since it does not require a large amount of cognitive effort, efficient since valence and arousal are registered with a single response, and easy to administer since a response merely requires pointing at the grid and clicking a button. Embedding the EmojiGrid in a VE may therefore enable users to report the affective appraisal of their experience in a minimally disruptive way.

In this study we evaluate the convergent validity of the EmojiGrid as an affective self-report tool for the appraisal of immersive VR experiences. Thereto, we measured the emotional response (valence and arousal) of 40 participants to 62 immersive 360° VR video clips from a database that has previously been validated (Li et al., 2017), and we compared our results with the corresponding ratings provided by the authors of this database. In addition, we obtained an objective measure of arousal by measuring Electro Dermal Activity (EDA), which has been established as a reliable marker for physiological arousal [Roth, 1983; see section Electrodermal Activity (EDA); Boucsein, 1999; Brouwer et al., 2018].

Background

Virtual reality (VR) can be defined as an interactive computer representation of a 3D environment with the ability to give the user a sense of immersion and presence (Gutierrez et al., 2008). The representation may either be a capture (video) of a real environment or a computer generated virtual environment (VE). A VR system allows the user to explore the represented three-dimensional space in real-time. “Immersion” refers to the technological capability of the system to encompass the user and isolate her from the real world (Slater and Wilbur, 1997). “Presence” is the subjective feeling of being located in a VE rather than in the place where one’s physical body is actually situated (see also Riva and Waterworth, 2003; e.g., Waterworth et al., 2015).

Immersive 360° VR videos are steadily growing in popularity on social media platforms (e.g., YouTube and Facebook). This is largely due to the increasing availability, quality and comfort of head-mounted displays (HMDs). An immersive 360° VR video is a photorealistic representation of a scene that updates with head-orientation. A total surround scene is obtained by stitching the recordings from multiple cameras together through software. In contrast to traditional videos, in which the user’s point of view is fixed and preset by the producer, a viewer watching a video in 360° VR format can interactively select at each instant the direction from which to see the VR scene, resulting in a more immersive experience (e.g., Ramalho and Chambel, 2013). Thus, the user has the freedom to explore the content based on her own interest. Some typical features distinguishing 360° VR video from regular videos are a wide field-of-view (the extent of the observable part of the scene) and natural control over the viewing direction.

Individuals using immersive VR systems that present a full 360° field of view can experience a significant degree of presence while showing a wide range of physiological and emotional responses (Riva et al., 2007; Felnhofer et al., 2015; Oliveira et al., 2018, 2020). The intensity of the elicited emotions is typically higher for more immersive VR systems (i.e., for systems that isolate the user’s senses from the external world) than for less immersive systems (Visch et al., 2010; Kim et al., 2014; Beck and Egger, 2018; Ding et al., 2018; Simon and Greitemeyer, 2019) and can be comparable to those evoked by real-life scenarios (Chirico and Gaggioli, 2019). While stereoscopic viewing may elicit stronger emotions (Peperkorn et al., 2015), this finding is not unequivocal (Baños et al., 2008; Ling et al., 2012) and probably depends on the VR contents and scenario. The interaction between presence and emotions appears to be mediated by arousal (Freeman et al., 2005; Diemer et al., 2015) and seems to be reciprocal: higher levels of presence induce stronger emotions (e.g., Västfjäll, 2003), while emotional VEs evoke higher levels of presence (Riva et al., 2007; Gorini et al., 2011). However, the level of presence in a VE seems to be more strongly influenced by emotional factors (i.e., by its content) than by technological attributes of the VE system (i.e., the degree of immersion). Thus, even low-immersive systems may induce high levels of presence if the VR scenario sufficiently engages emotions (Diemer et al., 2015).

The desire for emotional experiences is widely considered to be the main driver for using 360° VR in entertainment, underlying its increasing popularity on video streaming platforms.

Immersive VR is also applied in many other fields where emotions are an important factor, like studies involving consumer behavior and product evaluation (Sester et al., 2013; Bangcuayo et al., 2015; Bonetti et al., 2018; Andersen et al., 2019; Sinesio et al., 2019), the (participatory) design of (landscape) architecture (built) environments and soundscapes to assess the affective appraisal of planned environments (Mobach, 2008; Portman et al., 2015; Hayek et al., 2016; Echevarria Sanchez et al., 2017; Patterson et al., 2017; Puyana-Romero et al., 2017), and product design (Pitt et al., 2005; Söderman, 2005; Hilfert and König, 2016). Immersive VR is also used to study the

healing effects of natural environments and to provide restorative environments for relaxation purposes (Valtchanov et al., 2010; Calogiuri et al., 2018), such as the Sensiks Experience Pod (www.sensiks.com). The news industry has adopted immersive 360° videos (Hendriks Vettehen et al., 2019) to give viewers the feeling that they are present at the location of the event, so that they can experience the news story rather than merely watch it (de la Peña et al., 2010; Wang et al., 2018). By establishing an emotional connection between subject and viewer (Hendriks Vettehen et al., 2019) these videos may ultimately lead to greater audience's emotional involvement in current events (de la Peña et al., 2010). The tourism industry has embraced immersive VR systems as a powerful destination marketing tool (Guttentag, 2010; Huang et al., 2016; Marasco et al., 2018; Trindade et al., 2018). In addition to merely providing information, immersive VR systems can provide prospective travelers a compelling virtual experience of remote destinations by allowing them to explore real-world pre-captured (e.g., Fibbi et al., 2015; Tussyadiah et al., 2016) or real-time (e.g., drone-transmitted: Mirk and Hlavacs, 2014) 360° footage. Hence, these systems may provide customers a sneak preview or “*try before you buy*” travel pre-experience (Tussyadiah et al., 2016, 2017). This is expected to lead to better informed decisions and more realistic expectations, ultimately resulting in a more satisfactory vacation. In this type of experiential marketing (Schmitt, 1999), it is crucial to understand the viewer's emotional responses (Prayag et al., 2013; Beck and Egger, 2018), since travel-decisions are significantly influenced by momentary emotions (Walls et al., 2011). The goal is to enhance the persuasive power of VR by presenting destinations in creative ways that induce higher levels of arousal and positive valence (Tussyadiah et al., 2016).

Despite the increasing mass consumption of VE experiences in entertainment and the steadily growing number of people that watch VR movies using immersive HMDs, it is still not clear how the content of VR media relates to the user's emotional responses (Riva et al., 2007; Oliveira et al., 2018). Hence, more research is needed to fully understand how VR content and its underlying technology determines the emotions and sense of presence in VE (Seth et al., 2012), and particularly research that compares the emotional effects evoked by real-life experiences to those elicited by corresponding VR experiences (Riva et al., 2007). An essential requirement for these type of studies is the availability of validated and efficient tools for the assessment of affective responses to VR experiences (Oliveira et al., 2018). Preferably, these instruments should be able to unobtrusively blend into the VE so that users do not lose their sense of immersion when they need to step out of the VR to give a response (Regal et al., 2019; Krüger et al., 2020; Voigt-Antons et al., 2020). The recently introduced EmojiGrid is a viable candidate since it is intuitive and language independent (and therefore probably requires minimal cognitive effort: Kaneko et al., 2018b; Toet et al., 2018; Toet and van Erp, 2019; Voigt-Antons et al., 2020), and since it can easily be embedded in a VE.

Related Work

In previous studies on the emotional response to 360° VR systems, users reported their emotions using lists of emotional

terms (Suhaimi et al., 2018), verbal rating scales (Riva et al., 2007; Estupiñán et al., 2014; Sharar et al., 2016; Beck and Egger, 2018; Chirico and Gaggioli, 2019), the Self-Assessment Mannikin (SAM: Lang, 1980; e.g., Kim et al., 2014; Marín-Morales et al., 2018; Oliveira et al., 2018) or questionnaires like the Positive and Negative Affect Scale (PANAS: Watson et al., 1988; e.g., Macedonio et al., 2007; Riva et al., 2007; Ding et al., 2018) and the State-Trait Anxiety Inventory (STAI: Spielberger, 1983; e.g., Riva et al., 2007). These tools demand cognitive effort (interpretation) by the user and a significant amount of time to complete (e.g., the PANAS and STAI; Kaneko et al., 2018a) or successive cognitive interpretations of current sensations (Likert, VAS or SAM: Estupiñán et al., 2014; Sharar et al., 2016; Higuera-Trujillo et al., 2017; Marín-Morales et al., 2018; Oliveira et al., 2018). Since the use of these instruments is likely to affect the user's sense of presence, immersion and involvement, they are typically not embedded in the VE itself, but applied after ending the VR experience. As a result, the measurements thus obtained may not fully reflect the range of different emotions that were experienced over the entire course of the VE experience. This suggests a need for immersive affective self-report tools to assess VR evoked emotions, preferably at multiple instances during the VR experience itself (Bouchard et al., 2004; Regal et al., 2019; Oliveira et al., 2020). Recent studies showed that the inclusion of an affective rating tool in the VE can speed up user response by almost a factor of five compared to paper and pencil methods (Krüger et al., 2020), while the superposition of a two-dimensional affective response grid over 360° videos affords continuous affective ratings (Voigt-Antons et al., 2020; Xue et al., 2020). It has also been shown that self-report tools are less invasive and yield more reliable results when they used inside a VR, compared to their application outside the VR (Schwind et al., 2019; Putze et al., 2020).

The EmojiGrid

Although a range of explicit and implicit measures to measure a person's affective state is currently available, there is still no widely accepted method (Mauss and Robinson, 2009). Questionnaires are still the most practical method to assess emotions (Kaneko et al., 2018a). There are two types of questionnaires: verbal questionnaires (King and Meiselman, 2010; Spinelli et al., 2014; Nestrud et al., 2016) and graphical questionnaires (Bradley and Lang, 1994; Obaid et al., 2008; Vastenburger et al., 2011; Laurans and Desmet, 2012; Broekens and Brinkman, 2013; Huisman et al., 2013).

Verbal questionnaires enable people to report their affective state by rating or selecting words that most closely express their momentary affective state. However, these tools have several shortcomings (Toet et al., 2018): (1) people often find it difficult to find the right words to express their emotions, (2) in different cultures and languages emotions are described in different ways using different terms, (3) describing an experience as it happens can affect its nature, and (4) individuals have different vocabularies and language abilities. Also, it requires a considerable amount of time and cognitive effort to describe emotions in words (and this disadvantage increases when

questionnaires need to be filled out more than once over the course of an experiment).

Graphical tools enable users to report their affective state in a more intuitive and cognitively less demanding way by indicating the (part of the) figure that best represents their current feelings (see Zentner and Eerola, 2010 for a discussion of the advantages of graphical self-report tools; Toet et al., 2018). Instead of requiring users to verbalize their emotions, these tools use the human ability to intuitively and reliably link graphical elements to human emotions (Aronoff et al., 1988; Windhager et al., 2008; Larson et al., 2012; Watson et al., 2012).

A well-known and widely used graphical affective self-report is the SAM (Self-Assessment Mannikin; Bradley and Lang, 1994). Although the SAM has been validated in several studies, it has also been acknowledged that it has some serious drawbacks. First, although its graphical representation of the valence dimension appears quite intuitive, the way in which the dominance dimension is depicted appears much harder to understand, while the arousal dimension (depicted as an “explosion” in the stomach area) can be misinterpreted (Broekens and Brinkman, 2013; Betella and Verschure, 2016; Chen et al., 2018; Toet et al., 2018). Second, users need to respond to each of the affective dimensions separately and consecutively.

We therefore recently introduced the EmojiGrid (Figure 1; see also Toet et al., 2018) as an intuitive (language independent) and efficient alternative to the SAM. The EmojiGrid is a rectangular grid inspired by the Affect Grid (Russell et al., 1989), but instead of using text labels, its axes are labeled with emoji depicting iconic facial expressions. The expressions of the emoji along the horizontal (valence) axis range from *unhappy/not pleased* via *neutral* to *happy/pleased*, while their

intensity gradually increases along the vertical (arousal) axis. The opening of the mouth and the shape of the eyes represents the arousal dimension, while the concavity of the mouth, the orientation and curvature of the eyebrows, and the vertical distribution of these features over the area of the face corresponds to the degree of arousal. In contrast to the SAM, the EmojiGrid does not assess dominance, since the two-dimensional valence and arousal space that constitutes the so called “core affect” (Russell, 2003) is sufficient to describe basic emotions (Russell and Feldman Barrett, 1999; see also Gorini et al., 2009; Mattek et al., 2017).

Validation studies investigating both the individual emojis and their linear arrangement along the valence and arousal dimensions (i.e., the ordering of the emoji along the sides) of the EmojiGrid and their circular arrangement along its borders, showed that participants reliably interpreted the intended degrees of valence and arousal represented by the individual emoji on the EmojiGrid, and that the order in which the different emoji were arranged along the borders of the EmojiGrid appeared highly intuitive (Toet et al., 2018). In the context of food-evoked affective responses, we found that the EmojiGrid yielded appropriate responses whereas the SAM’s arousal dimension was typically misinterpreted (Toet et al., 2018). The EmojiGrid has been extensively validated with different age groups, cultures and ethnicities (Kaneko et al., 2018b; Toet et al., 2018, 2019b; Toet and van Erp, 2019).

Users can report their subjectively experienced valence and arousal through the EmojiGrid by positioning a cursor on the location that best represents their affective appraisal and by pressing a response button to register their response (e.g., by pointing and clicking with a mouse). A simple linear scaling can be applied to map the two dimensions of the EmojiGrid to the desired response range. The tool is easy to use, requires minimal instructions (the facial expressions are intuitive and do not need additional explaining), and is efficient (a single response suffices to report both valence and arousal). This suggests that embedding the EmojiGrid in a VE could afford users a way to report their affective response while minimally disrupting the VR experience itself.

Electrodermal Activity (EDA)

Electrodermal activity (EDA) mainly results from an activation of the sweat glands (Andreassi, 2013). Sweat production increases the conductivity of current through the skin. Sweat glands are directly linked to the “fight or flight” sympathetic branch of the autonomous nervous system (Dawson et al., 2000; Benedek and Kaernbach, 2010) such that EDA is a good measure of physiological arousal. Indeed, EDA has been established as a reliable marker for physiological arousal across different contexts (Roth, 1983; Boucsein, 1999; Brouwer et al., 2018; Kaneko et al., 2019) and cannot be affected by demand characteristics or other subjective response biases that may affect self-report (Higuera-Trujillo et al., 2017).

Current Study

In this study the EmojiGrid was used to obtain subjective ratings of valence and arousal for immersive 360° VR video

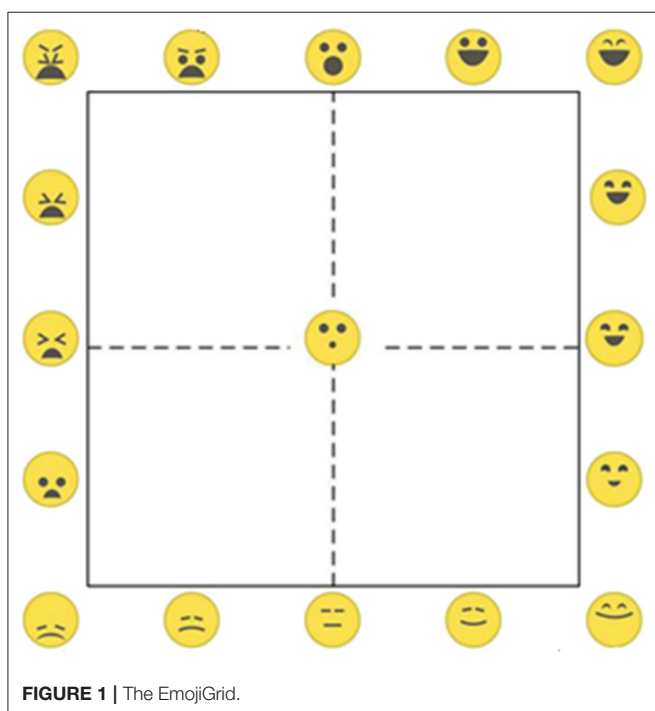


FIGURE 1 | The EmojiGrid.

clips from a validated database (Li et al., 2017). The results were compared with corresponding ratings (provided with the database) obtained with the validated SAM self-report tool. Also, simultaneous EDA measurements were obtained as an objective measure of physiological arousal.

METHODS

In this study, participants viewed 360° VR movies using a HMD. They used an EmojiGrid that was embedded in the VE to report their subjective affective response (valence and arousal) to each movie. Their autonomic physiological response (arousal) to the movies was measured through electrodermal activity.

Participants

A total of 40 Caucasian participants (22 females, 18 males) aged between 18 and 29 ($M = 22.16$; $SD = 2.70$) took part in this study. Participants were recruited either via direct approach or from the TNO pool of volunteers for human-subject experiments. Exclusion criteria were age (outside the range of 18–35 years old) and hearing or (color) vision deficiencies. Age was an exclusion criterion since recent research has shown that older adults tend to be more susceptible to cybersickness than younger ones (Arns and Cerney, 2005; Knight and Arns, 2006; Park et al., 2006; Liu and Uang, 2007). Hearing and vision deficiencies were adopted as an exclusion criterion since these limitations may influence the affective experience of the VR videos. Physical (heart diseases, high or low blood pressure, epilepsy, etc.) or mental (phobias like claustrophobia, acrophobia, arachnophobia, etc.) health issues were also exclusion criteria since people suffering from these issues might experience problems while watching some of the affectively intense immersive VR videos used in this study. Of the total of 40 participants, 21 had previous experience with watching immersive VR videos on HMDs, while 19 had no previous experience.

The experimental protocol was reviewed and approved by the TNO Internal Review Board (IRB) (Approval Ref: 2019-024) and was in accordance with the Helsinki Declaration of 1975, as revised in 2013 (World Medical Association, 2013). Participation was voluntary. All participants received a financial compensation for their participation (€20 for participants that were directly approached, while participants from the TNO pool received €10 extra to reimburse their additional travel expenses).

Stimulus Material

The stimuli used in this study were 62 immersive VR videos from the public database composed by Li et al. (2017). The videos originate from sources like YouTube, Vrideo, and Facebook. All videos are of short length (<12 min), require no explanation, and have been found to induce different levels of valence and arousal (Li et al., 2017). The database provides mean valence and arousal ratings for each video, as established with the Self-Assessment Mannikin (SAM; Lang, 1980). **Table 1** shows a list of all the clips used in this study, together with their identifiers and title in the original database provided by Li et al. (2017), their length (s), their corresponding valence and arousal ratings measured both

with the SAM by Li et al. (2017) and with the EmojiGrid (this study), and the EDA arousal measures.

Although the original database consists of 73 videos, not all videos could be included in this study. Some videos (nrs. 9, 11, 29, 36, 45) were no longer available online, while others (nrs. 17, 31, 51, 55, 59, 61) were not in a usable format. As a result, 11 videos could not be included. The duration of the remaining 62 videos varied from 37 to 668 s with an average of 187 s per video. The arousal ratings varied from 1.57 to 7.42 on a scale from 1 to 9 ($M = 4.32$, $SD = 1.41$), which is comparable to the arousal ratings of the IAPS images from the standard International Affective Picture System (IAPS; Lang et al., 2005), which range from 1.72 to 7.35. The valence ratings of the VR clips varied from 2.18 to 7.77 ($M = 5.56$, $SD = 1.44$), which also agrees reasonably well with the valence ratings of the IAPS images, which range from 1.31 to 8.34. All participants watched a subset of the immersive VR videos, such that each video was seen and evaluated by at least seven participants.

Li et al. (2017) found that some participants became nauseous and/or fatigued after uninterrupted watching of immersive 360° VR videos for more than 15 min, while most were at ease with a duration of about 12 min. We therefore divided the 62 immersive 360° VR videos used in this study into clusters with a duration of about 12 min per cluster. The result was a set of 16 clusters, with two to six videos per cluster (see **Table 2**). Also, following Li et al. (2017) procedure, at most two clips of a particular valence (negative/positive) or arousal (low/high) were shown consecutively. Positive and negative valenced videos in the same subsets were presented in random order (i.e., some subsets started with a negative video, while others started with a positive or neutral video). The order in which the clusters were presented to the participants was also randomized. An Excel file listing the links to the original video clips and the experimental details and results for all videos that were used in this study is provided in the **Supplementary Material**. All stimuli are available from the OSF repository at <https://osf.io/9qgce> with doi: 10.17605/OSF.IO/9QGCE.

Stimulus Presentation

The immersive monoscopic VR videos were presented on a Samsung Odyssey Windows Mixed Reality headset (www.samsung.com), equipped with a Dual 3.5" AMOLED 1440 × 1600 resolution display, a 110° field of view and a refresh rate of 90 Hz. The low-latency tracking technology of the Odyssey HMD determines the relative position of the viewer's head and adjusts the view of the immersive video accordingly. To prevent any distractions by ambient sound, the video soundtracks were presented through a Sony MDR-1000× noise-canceling headset.

The EmojiGrid was embedded in the virtual environment at the end of each video using the A-Frame (<https://aframe.io>) open source Javascript framework for creating (web-based) VR experiences. Using Node.js (<https://nodejs.org>), a local server was set up on an Alienware 13 R3 Notebook (Intel Core i7 7700HQ) which ran on Windows 10. Participants used a Samsung Odyssey remote control to point a graphical raycast beam and place a check mark at the appropriate location on the EmojiGrid when rating the videos (**Figure 2**). The EmojiGrid was displayed with a

TABLE 1 | List of the 62 videoclips used in this study, with their identifiers (ID) and titles from the original database by Li et al. (2017), their lengths (s), the corresponding valence and arousal values measured by Li et al. (2017) using the SAM (V_sam, A_sam), and with the EmojiGrid V_eg, A_eg: this study, and the EDA values (μ S).

Item	ID	Title	Length (s)	V_sam	A_sam	V_eg	SD_V_eg	A_eg	SD_A_eg	EDA (μ S)
1	1	Abandoned building	120	4.39	2.77	5.46	0.42	2.72	0.38	0.46
2	2	A Mumbai Summer	199	5.87	4.60	4.81	0.30	4.97	0.47	0.52
3	3	Abandoned City	50	3.33	3.33	4.88	0.59	3.45	0.76	0.26
4	4	Jared Leto Tour Guides Alaska's Melting Glaciers	234	4.73	3.33	3.43	0.50	4.04	0.80	0.42
5	5	Chernobyl VR 360	548	3.06	4.18	2.85	0.59	3.26	0.73	0.84
6	6	Sadness Elicitation – Lake Valley	119	5.36	2.64	6.71	0.44	3.14	0.43	0.71
7	7	Fukushima	560	2.69	4.63	2.10	0.23	3.21	0.60	0.49
8	8	Happyland 360	611	3.33	3.40	3.74	0.58	3.19	0.61	0.25
9	10	New York 2121	120	4.00	3.93	5.26	0.49	4.65	0.65	0.89
10	12	The fight to save threatened species	124	7.00	4.60	6.70	0.59	5.13	0.76	0.71
11	13	The Margins	137	4.92	4.08	3.13	0.36	5.44	0.66	0.63
12	14	War Zone	183	2.53	3.82	2.12	0.29	4.55	0.56	0.47
13	15	Inside a bee hive	43	3.69	3.94	4.31	0.74	6.03	0.69	1.26
14	16	Solitary Confinement	221	2.38	4.25	2.86	0.27	4.47	0.70	0.84
15	18	The Displaced	668	2.18	4.73	2.81	0.20	3.12	0.60	0.31
16	19	The Nepal Earthquake Aftermath	240	2.73	3.80	2.68	0.29	3.44	0.43	0.50
17	20	War Knows No Nation	448	4.93	6.07	5.54	0.49	6.12	0.75	0.45
18	21	Zombie Apocalypse Horror	265	3.20	5.60	3.12	0.40	6.21	0.43	0.83
19	22	Great Ocean Road	118	7.77	3.92	7.64	0.22	4.50	0.51	0.69
20	23	Instant Caribbean Vacation	150	7.20	3.20	7.10	0.33	3.83	0.65	0.57
21	24	Blyde Canyon	157	4.82	3.09	6.53	0.39	3.04	0.60	0.56
22	25	The Most Beautiful Place in the World	186	6.65	4.94	6.88	0.32	4.69	0.77	0.66
23	26	Getting Licked by a Cow in Ireland	65	7.07	3.21	6.61	0.41	5.35	0.67	0.62
24	27	Seagulls	120	6.00	1.60	5.39	0.44	2.63	0.23	0.60
25	28	Maldives beach and resort	138	6.69	3.50	7.16	0.39	3.98	0.52	0.42
26	30	Haleakala National Park Sunrise	37	6.72	3.39	7.46	0.25	5.04	0.63	0.76
27	32	Malaekahana Sunrise	120	6.57	1.57	7.07	0.36	2.44	0.43	0.71
28	33	Pacific Sunset Half Moon Bay	134	6.19	1.81	7.42	0.35	3.35	0.64	0.62
29	34	Raising Ducklings	203	6.00	2.63	7.21	0.43	4.05	0.33	0.42
30	35	Redwoods Walk Among Giants	120	5.79	2.00	7.52	0.45	3.98	0.60	0.62
31	37	Sunset of Oia-Santorini	89	6.55	3.09	6.10	0.52	3.50	0.46	0.45
32	38	Mountain Stillness	128	6.13	1.80	7.72	0.16	4.02	0.65	0.91
33	39	Zip-lining in Chattanooga	127	4.79	4.57	6.41	0.38	5.92	0.61	1.41
34	40	VRKittens	101	6.07	4.00	7.93	0.26	6.38	0.41	0.90
35	41	Fighter Jet Patrouille Suisse	120	6.55	4.73	7.24	0.30	6.66	0.71	0.47
36	42	Cute Kittens Battle	65	6.94	4.13	8.31	0.28	5.08	0.81	0.52
37	43	ALICE the first Swedish baby goes VR	126	7.33	3.44	7.07	0.41	5.67	0.71	0.85

(Continued)

TABLE 1 | Continued

Item	ID	Title	Length (s)	V_sam	A_sam	V_eg	SD_V_eg	A_eg	SD_A_eg	EDA (μ S)
38	44	Conquer the Mega Ramp	86	5.29	6.43	5.53	0.15	6.65	0.42	1.61
39	46	Explore the world with IM360	197	6.59	4.29	7.77	0.22	4.62	0.67	0.55
40	47	Puppy Bowl XII	192	7.44	4.75	7.11	0.45	4.63	0.66	0.47
41	48	Holi Festival of Colors	173	6.60	4.00	7.04	0.23	4.63	0.54	0.61
42	49	India's first ever 360 Wedding Video	201	7.07	4.00	6.51	0.30	4.60	0.49	0.60
43	50	Puppies host SourceFed for a day	80	7.47	5.35	7.35	0.30	6.80	0.39	0.55
44	52	Speed Flying	154	6.75	7.42	7.12	0.36	7.13	0.54	0.60
45	53	Tomorrowland 2014	265	5.80	5.40	6.32	0.39	6.54	0.39	0.70
46	54	As It Is	154	7.00	4.67	7.16	0.32	7.09	0.44	0.60
47	56	Solar Impulse assembles the Mobile Hangar	129	5.80	3.80	4.99	0.25	4.08	0.35	0.39
48	57	Les berges du centre à Wasquehal	87	5.75	3.25	6.19	0.26	4.56	0.52	0.51
49	58	Spangler Lawn	58	5.09	3.27	5.41	0.37	2.79	0.52	0.25
50	60	Russian Knights acrobatic rehearsals	120	5.73	4.20	6.57	0.32	6.32	0.60	0.75
51	62	Mega Coaster	117	6.17	7.17	6.10	0.53	7.78	0.21	1.44
52	63	NASA: Encapsulation and Launch of OSIRIS Rex	285	6.36	5.93	6.70	0.25	5.82	0.60	0.74
53	64	Surrounded by elephants	156	5.94	5.56	6.78	0.47	4.75	0.51	0.45
54	65	Kidnapped	406	4.83	5.25	5.84	0.61	5.36	0.81	0.39
55	66	Great Hammerhead Shark Encounter	134	6.17	6.67	4.59	0.56	6.77	0.65	0.53
56	67	Canyon Swing	104	5.38	6.88	5.03	0.55	6.41	0.37	1.13
57	68	Jailbreak 360	339	4.40	6.70	6.74	0.45	5.99	0.66	0.51
58	69	Walk the tight rope	151	6.46	6.91	5.22	0.60	5.41	0.71	0.93
59	70	Tahiti Surf	205	7.10	4.80	7.72	0.24	6.71	0.48	0.67
60	71	Lion's Last Stand	40	5.88	5.25	6.41	0.37	3.90	0.79	0.80
61	72	Relive Undertaker's Entrance	122	5.36	5.57	5.07	0.53	3.56	0.43	0.38
62	73	Through Mowgli's Eyes	93	6.27	6.18	7.58	0.25	6.47	0.46	1.09

diameter with an angular size of 31 degrees, while the diameter of the individual emojis was 2.2 degrees.

Timestamps marking the start and end of each video, were logged in EPOCH format using A-Frame and served to calculate mean EDA responses [see section Electrodermal Activity (EDA)] over the runtime of each individual video.

A-Frame was also used to log the coordinates of the check marks (subjective responses) on the EmojiGrid. As the location of the EmojiGrid was fixed in the VE, a check mark's position on the grid could be determined from its coordinates. These coordinates were rounded to two decimal places.

Measures

Demographics

Participants were asked to report their age and gender.

Subjective Valence and Arousal: EmojiGrid

Participants rated valence and arousal for each VR video clip by pointing a graphical raycast beam at the location of the EmojiGrid (Figure 1; see also Toet et al., 2018) that best represented their affective appraisal of the 360° VR video they had just seen (Figure 3). The responses were digitized with 12 bit accuracy and mapped to a response scale ranging between 1 and 9, to enable comparison with the corresponding SAM ratings provided by Li et al. (2017).

Electrodermal Activity (EDA)

Electrodermal activity (EDA) was recorded with the EdaMove 4 EDA and Activity Sensor (Movisens GmbH, Karlsruhe, Germany). Two self-adhesive, gelled electrodes (Ag/AgCl, MTG 102 IMIELLA electrode, W55 SG, textured fleece electrodes,

55 mm diameter) were fixed on the palmar surface of the participant's left hand. In addition, skin-friendly adhesive fleece tape was applied to ensure the attachment of the electrodes. The device applied a constant DC voltage of 0.5 V to the skin. The EDA signal was sampled at 32 Hz. This signal is composed of two components: a tonic (slow) and phasic (fast) activity component (Boucsein et al., 2012). Phasic activity (also referred to as skin conductance response or SCR) most clearly reflects responses to arousing events (e.g., Brouwer et al., 2018). In this study we determined the mean phasic activity (SCR) for each participant and for each immersive VR video over the entire duration of the video.

Procedure

After arriving at the test location, the experimenter welcomed the participants and gave them a verbal introduction and

TABLE 2 | Composition (clip IDs) and total length (s) for each of the 16 clusters.

Cluster	Video clips	Total length (s)
1	2, 6, 19, 30, 35	715
2	5, 48	721
3	7, 54	714
4	8, 67	715
5	12, 13, 25, 41, 52	721
6	16, 27, 63, 73	719
7	18, 58	726
8	20, 21	713
9	14, 22, 23, 26, 49	717
10	1, 15, 28, 33, 34, 37	727
11	10, 32, 38, 39, 40, 43	722
12	3, 44, 46, 47, 58, 66	717
13	24, 50, 53, 56, 57	718
14	4, 60, 62, 70, 71	716
15	64, 65, 69	713
16	42, 47, 68, 72	718

instructions. Participants were informed that they would be presented with immersive VR videos and were asked to rate their affective appraisal of each video. Then, they read and signed the IRB approved informed consent form and filled in a pre-questionnaire about demographic characteristics.

Next, a printed copy of the EmojiGrid was presented (**Figure 1**). Participants were told they could use this tool to report their emotional response to a video by clicking on a point in the grid. To get familiar with the EmojiGrid, they were given (a maximum of) 2 min to inspect it. No reference was made to the concepts of valence and arousal (the constructs underlying the axes of the EmojiGrid), since we wanted the participants to use the tool intuitively.

Then, the physiological measurements were explained, and the participants were fitted with the EDA sensors. To obtain a baseline for the EDA measurements, the participants were first asked to sit calmly and relax for 1 min.

Next, participants were introduced to the Samsung Odyssey HMD. After putting on this device, they watched and rated a practice cluster consisting of three immersive VR videos from YouTube, with respectively low (broiler chickens on a factory farm: <https://www.youtube.com/watch?v=AfXjCJcexSI>), moderate (a walk through a park: <https://www.youtube.com/watch?v=WXqV7OcbnPY>), and high (panda recess at the Wolong National Nature Reserve in China: https://www.youtube.com/watch?v=Jc7mqsD_sWM) valence. Participants were told that they should feel free to look around in the 360° VR scene. This introduction accustomed the participants to the experimental procedure. After this practice trial, the actual experiment started.

During the experiment, each participant was presented with three clusters of immersive 360° VR videos. After watching each video, the participants were asked to indicate the valence and arousal components of their emotional state by pointing the graphical raycast beam at the EmojiGrid using the Samsung Odyssey Controller. By clicking the response button of the

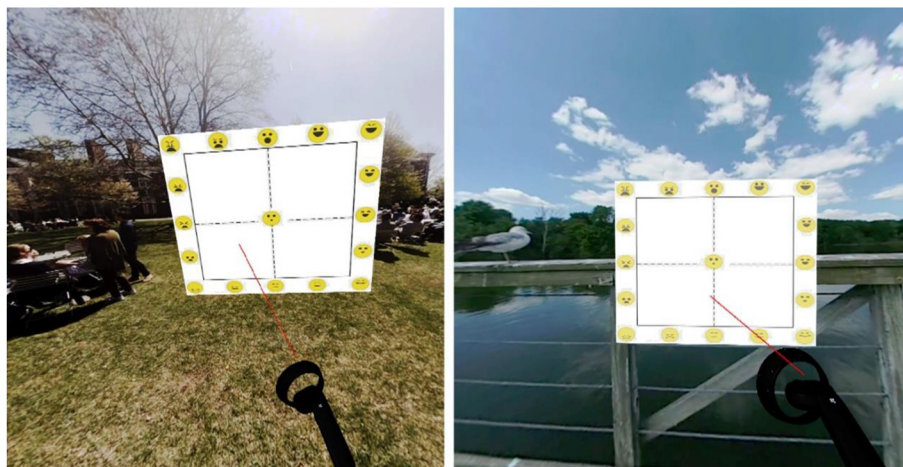


FIGURE 2 | Screenshots from two different 360 VR clips, showing the Samsung Odyssey Controller (bottom right) projecting a graphical raycast beam on the EmojiGrid that is embedded in the VR scene.

controller, their response (i.e., the coordinates of the check mark on the EmojiGrid) was registered, and the next video in the cluster was started. The participants were allowed to take a 5 min break between the presentation of two clusters. They could remove their HMD during the breaks.

After finishing the viewing and rating process of the third (last) cluster, both the HMD and the EDA sensors were removed, and the participants were debriefed about the purpose of the study. The entire experiment lasted about 80 min. A timeline of the entire experimental procedure is shown in **Figure 3**.

To avoid any distractions, the experiments were performed in a quiet room. Watching immersive VR videos on head mounted

displays is known to induce postural instability (Munafò et al., 2017). In this study the participants were therefore seated on a swivel chair during the entire VR experience. This enabled them to look around and take in the full 360° virtual environment in a safe manner.

Data Processing and Analysis

For each VR video we computed the mean valence and arousal responses over all participants. Matlab 2019a (www.mathworks.com) was used to plot the data and to compute a least-squared linear mapping between the valence and arousal tuples measured with (a) the EmojiGrid in this study and (b) with the SAM in the study by Li et al. (2017).

Visual checks were performed on all collected EDA signals to identify and remove erroneous data (e.g., resulting from technical problems such as signal dropout due to loose contacts). The remaining EDA data of 37 participants was analyzed. Phasic activity (SCR) was extracted using Continuous Decomposition Analysis (CDA) as implemented in Ledalab (www.ledalab.de; Benedek and Kaernbach, 2010), and was averaged across individual video epochs.

IBM SPSS Statistics 25 (www.ibm.com) for Windows was used to perform all statistical analyses. To test the convergent validity of the EmojiGrid we computed intraclass correlation coefficients (ICCs) between the mean valence and arousal ratings measured with (a) the EmojiGrid in this study and with (b) the SAM by Li et al. (2017). The ICC reflects not only the degree of correlation but also the agreement between the different measurements (the reliability of the method: Hallgren, 2012; Koo and Li, 2016). ICC estimates and their 95% confident intervals were based on a mean-rating ($k = 3$), consistency, 2-way mixed-effects model (Shrout and Fleiss, 1979; Hallgren, 2012; Koo and Li, 2016). ICC values < 0.5 are indicative of poor reliability, values between 0.5 and 0.75 indicate moderate reliability, values between 0.75 and 0.9 indicate good reliability, while values > 0.9 indicate excellent reliability (Cicchetti, 1994). For all other analyses a probability level of $p < 0.05$ was considered to be statistically significant. Cohen's d was used to quantify the significance of the difference in the mean phasic activity scores between videos that were subjectively scored as high and low arousing.

RESULTS

Inspection of the self-report and physiological data both clearly indicate that the 360° VR videos successfully induced a range of emotions, differing widely in valence and arousal. For instance, participants' arousal varied across the different video clips as expected. Videos showing hang gliding (#52: Speed Flying) and roller coaster (#62: Mega Coaster) experiences were rated highest on arousal, while videos showing birds (#27: Seagulls) and a sunrise beach (#32: Malaekahana Sunrise) were rated lowest on arousal. Videos showing beaches (#22: Great Ocean Road) and puppies (#50: Puppies host SourceFed for a day) were rated highest on valence, while videos showing an isolation prison cell (#16: Solitary Confinement) and war refugees (#18: The Displaced) received the lowest valence ratings. Also, videos subjectively rated as most arousing also evoked

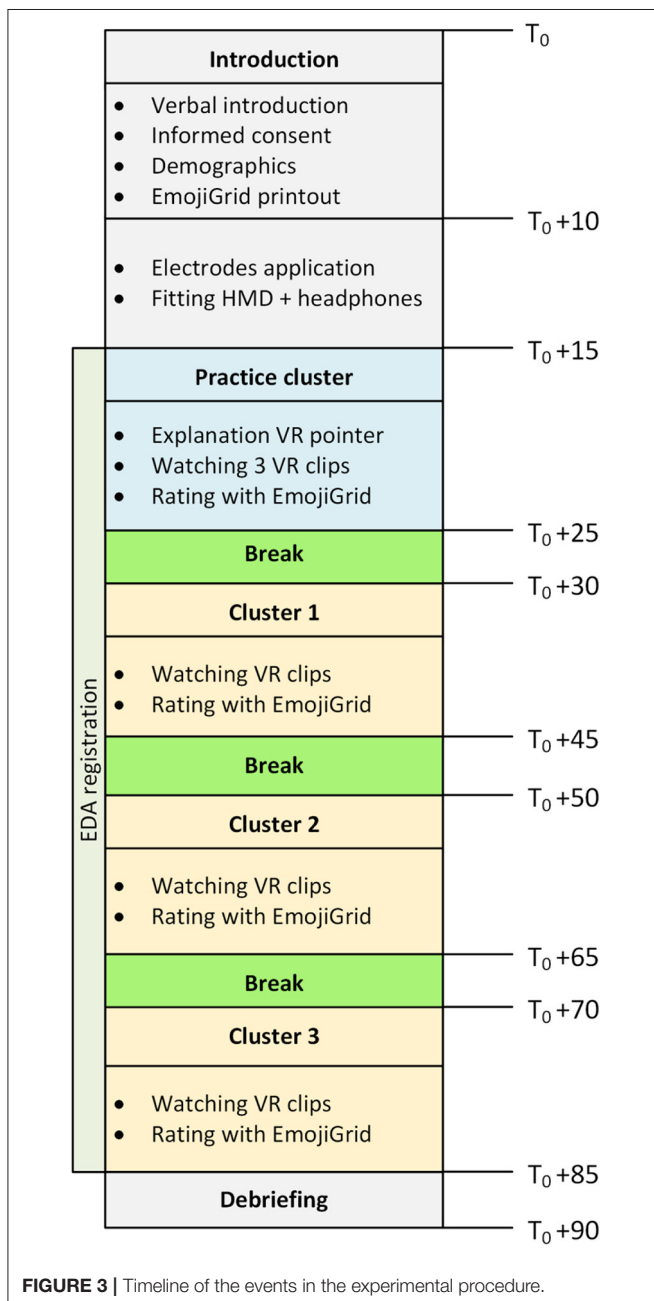


FIGURE 3 | Timeline of the events in the experimental procedure.

the highest EDA levels, while videos subjectively rated as least arousing also evoked the lowest EDA levels (see section Electrodermal Activity).

Valence and Arousal Ratings

For each video, the mean and standard deviation response for valence and arousal across all viewers was computed. **Figure 4** shows the relation between the mean subjective valence and arousal ratings obtained with the EmojiGrid in the current study and the corresponding ratings obtained with the SAM by Li et al. (2017). To quantify the agreement between both results we computed the Intraclass Correlation Coefficient (ICC) estimates and their 95% confidence intervals, based on a mean-rating, consistency, two-way mixed-effects model.

The ICC for valence was 0.91 (with a 95% confidence interval ranging between 0.85 and 0.95) and the ICC for arousal was 0.83 (with a 95% confidence interval ranging between 0.72 and 0.90). Thus, the valence ratings obtained by both studies are in excellent agreement, while the arousal ratings are in good agreement.

Electrodermal Activity

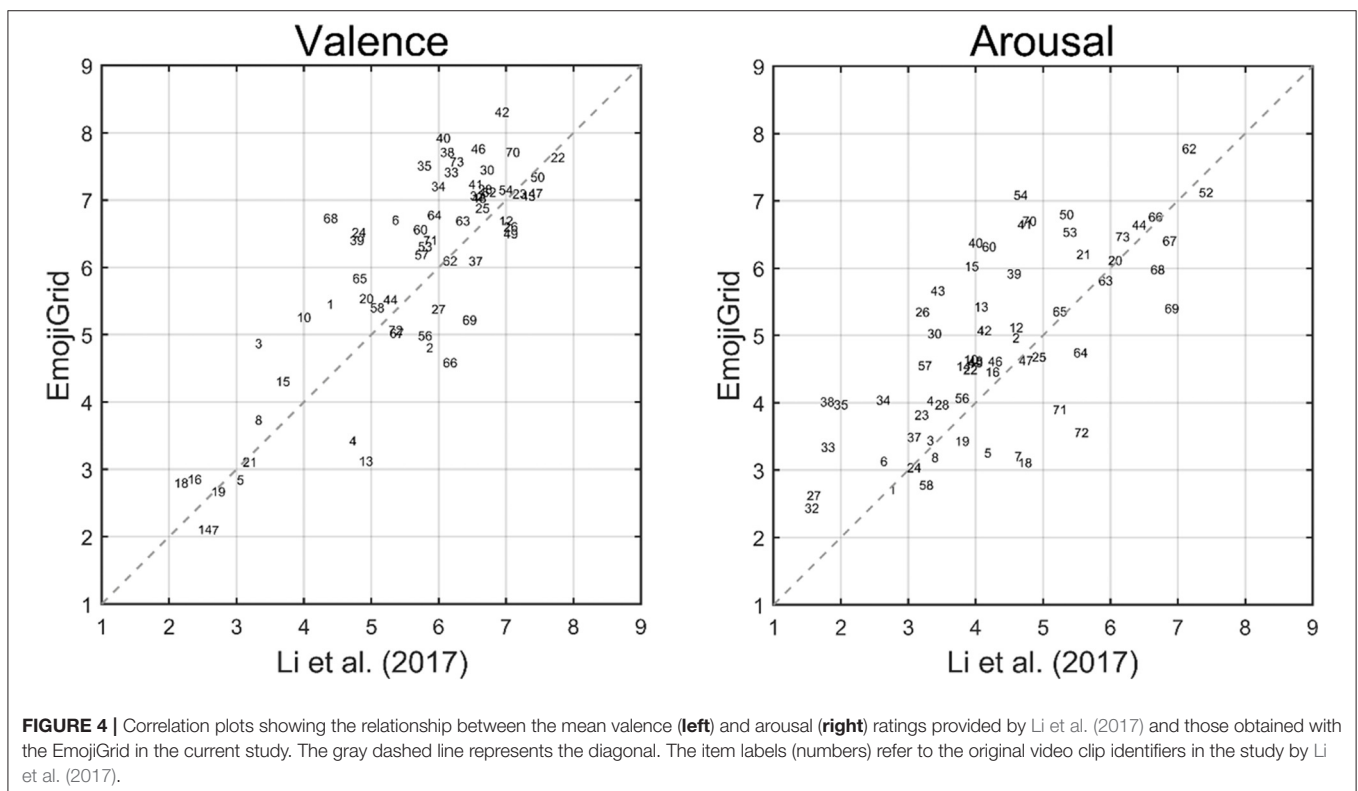
To examine the relation between the EDA signal and the subjective arousal ratings we computed the mean of the phasic activity for each video and across all viewers. The Pearson product-moment correlation between the mean EDA values and the self-reported arousal ratings obtained with the EmojiGrid was 0.49 ($p < 0.0001$), and the correlation with the SAM ratings from the study by Li et al. (2017) was 0.30 ($p < 0.02$). This indicates that

videos that are subjectively rated as more arousing also evoked higher levels of physiological arousal.

To further investigate this relation, we determined for each individual participant the three most arousing and the three least arousing VR videos, using their self-reported arousal ratings obtained with the EmojiGrid. The mean phasic activity values corresponding with these videos were then used to calculate the overall average phasic activity over the three most and the three least arousing videos. This way, two mean phasic activity values (one corresponding to the three most arousing videos and one for the three least arousing videos) were obtained for each participant. A paired-samples t -test was conducted to evaluate the difference in mean phasic activity scores for immersive VR videos that were subjectively rated as high and low arousing. There was a statistically significant difference in mean phasic activity scores between videos that were subjectively scored as high arousing ($M = 0.74 \mu\text{S}$, $SD = 0.63$) and low arousing ($M = 0.56 \mu\text{S}$, $SD = 0.52$), $t(36) = 3.54$, $p < 0.001$ (two-tailed). The mean difference in mean phasic activity scores was $0.18 \mu\text{S}$ with a 95% confidence interval ranging from 0.08 to $0.28 \mu\text{S}$. Cohen's d was 0.65, indicating a medium effect size. This indicates that the immersive VR videos subjectively rated as more arousing on the EmojiGrid indeed evoked higher levels of physiological arousal.

DISCUSSION AND CONCLUSIONS

We found that participants were able to intuitively use the EmojiGrid without any further explanation of the concepts of



valence and arousal. They efficiently rated both valence and arousal by pointing a raycast beam onto an EmojiGrid that was embedded in the immersive VE and clicking a response button. These are obvious benefits compared to for instance the popular SAM tool, that has an arousal scale that can be misunderstood (Broekens and Brinkman, 2013; Betella and Verschure, 2016; Chen et al., 2018; Toet et al., 2018) and which requires two successive ratings on separate scales.

The first objective of this study was to investigate the convergent validity and effectiveness of the EmojiGrid for the affective appraisal of immersive 360° VR video clips. It appeared that participants could effectively report the degrees of valence and arousal associated with the emotion they experienced after seeing an immersive VR video by simply pointing at an EmojiGrid that was embedded in the VE. To assess the convergent validity of the EmojiGrid for measuring the valence and arousal of immersive VR videos, we compared (1) the mean subjective valence and arousal ratings for all videos obtained with this tool to (2) the corresponding ratings from the study by Li et al. (2017) that were obtained with the validated SAM rating tool. The agreement between both studies was excellent for the subjective valence ratings and good for the subjective arousal ratings. The arousal ratings obtained with the EmojiGrid also corresponded well with another validated and objective arousal standard, namely EDA. Moreover, the correlation between subjective and objective arousal ratings was stronger for the EmojiGrid than for the SAM. Thus, it appears that the EmojiGrid may be used as a language-independent and efficient self-report tool for the affective appraisal of VE-induced emotions. Compared to most existing methods, the EmojiGrid is more efficient (both valence and arousal are rated with a single response) and needs no verbal labels or extensive explanation (it is intuitive and language independent).

The second goal of this study was an objective validation of the arousal scale of the EmojiGrid by relating subjective arousal ratings to electrodermal measurements. Our results show that immersive VR videos that were subjectively rated on the EmojiGrid as more arousing also induced higher levels of physiological arousal. This indicates that subjective ratings of arousal obtained with the EmojiGrid reliably reflect objective autonomic physiological arousal.

The present results suggest that the EmojiGrid may also be used for the real-time affective evaluation of events in VR or for giving affective feedback. For instance, in studies on affective communication in human-computer interaction (e.g., Tajadura-Jiménez and Västfjäll, 2008), the EmojiGrid could be used to enable users to repeatedly report perceived affect (e.g., by projecting a pointer-controlled beam on the grid). Such an application could also be useful for the affective annotation of multimedia (Chen et al., 2007; Soleymani et al., 2008; Runge et al., 2016; Suhaimi et al., 2018), for personalized affective multimedia retrieval (e.g., through query-by-emoji: Cappallo et al., 2019) or multimedia recommender systems (Hanjalic and Xu, 2005; Xu et al., 2008; Zhang et al., 2010; Lopatovska and Arapakis, 2011; Koelstra et al., 2012; Zhao et al., 2013), for the affective appraisal of multimedia entertainment in real-time (Fleureau et al., 2012), to give affective feedback in serious gaming applications (Anolli

et al., 2010), and for affective multimedia generation (e.g., music: Kim and André, 2004). We are currently implementing the EmojiGrid in a multisensory VR environment (the Sensiks Sensory Reality Pod: www.sensiks.com) as an interface for the user to select and adjust the desired multisensory (visual, auditory, tactile, and olfactory) affective experiences.

Previous studies suggest that the EmojiGrid is more intuitive and therefore may require less cognitive effort than other affective rating tools, making it more efficient to use. To test this hypothesis, future studies should compare the EmojiGrid with different forms of traditional affective rating tools (e.g., a label-based valence-arousal grid or SAM) embedded in a VE.

Although users can report both valence and arousal with a single click, the EmojiGrid still requires a successive judgement along both axes. Future studies should investigate whether the superposition of additional emoji over the inner area of the grid (with facial expressions “interpolating” those of the emoji near its outer edges) can further improve the intuitiveness of this tool.

LIMITATIONS OF THE PRESENT STUDY

A limitation of the present study (and also of the study by Li et al., 2017) was the fact that we did not register the viewing direction and did not check what trigger events in the VR movies the participants observed or missed. Although we found no outliers in the current data, a variation in viewing direction may have caused a variation in the obtained assessments, because the participants may have missed some relevant events. Also, participants reported that they found it difficult to give an overall rating to videos containing both pleasant and unpleasant episodes. Future studies should therefore (1) track the instantaneous viewing direction to check what parts of the scene or which events participants actually perceived, and (2) afford a more frequent rating procedure to ensure that participants can rate affectively distinct parts of a VR movies individually, thus eliminating the need to give an overall rating to videos containing episodes of opposite valence. This can for instance be achieved by continuously blending the EmojiGrid in the field-of-view of the observer (e.g., Voigt-Antons et al., 2020) or placing it as a billboard at a fixed location in the scene (e.g., Regal et al., 2019).

Another limitation of this study is that we did not measure the occurrence of cybersickness or discomfort during the experiments. During the debriefing the participants reported that they had not experienced any signs of nauseousness or fatigue after uninterrupted watching the blocks of immersive 360° VR videos for about 12 min. However, this does not preclude that they unconsciously experienced a low degree of cybersickness that influenced their affective state. This could in turn affect the present results, since current mood is known to have a direct impact on subsequent judgements through misattribution (Schwarz and Clore, 1983; Schwarz, 2002; Russell, 2003; Clore and Huntsinger, 2007).

Another limitation of this study is the fact that some of the VR videos were rather long and reflected different emotion overtime, while the participants had to evaluate their emotional response in a summative fashion after watching the whole

clip. However, this also applies to the original study by Li et al. (2017), who's experimental protocol we closely followed in this study to guarantee that the results of both studies are comparable.

In this study we did not measure the valence polarity of the user's experience from physiological data. Hence, a full objective validation of the valence scale of the EmojiGrid is still lacking. Future studies using the EmojiGrid for measuring the affective responses to immersive VR experiences should include physiological measures that correlate with subjective valence ratings, such as facial electromyograph (EMG) activity (Mavridou et al., 2018).

DATA AVAILABILITY STATEMENT

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

ETHICS STATEMENT

Written informed consent was obtained from the individuals for the publication of any potentially identifiable images included in this article.

AUTHOR CONTRIBUTIONS

AT had the original idea and supervised this study, provided the resources, analyzed, visualized and curated the data, wrote the original draft paper, and prepared the final manuscript. FH performed the experiments, collected the data, performed the formal analysis, and wrote the original draft paper. A-MB

assisted with the original draft preparation and critically revised the final manuscript on article structure, logical organization, and language. TM assisted with the original draft preparation and critically revised the final manuscript on article structure, logical organization, and language. JE critically revised the final manuscript on article structure, logical organization, and language. All authors contributed to the article and approved the submitted version.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: <https://www.frontiersin.org/articles/10.3389/fvrv.2020.552587/full#supplementary-material>

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Conflict of Interest: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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