Cue-reactivity in behavioral addictions: A meta-analysis and methodological considerations

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INTRODUCTION

Persons who are addicted to substances frequently show specific responses when they are confronted with substance-relevant cues. Responses include subjective, peripheral physiological, and neural responses. Carter and Tiffany (1999) published a meta-analysis including persons addicted to alcohol, nicotine, cocaine, or heroin. Subjective and peripheral physiological responses, such as heart rate, sweat gland activity, and skin temperature of the addicted persons, were analyzed. These reactions were compared between addiction-relevant cues versus control cues. Results indicate that addiction-relevant cues elicited cue-reactivity on a subjective, and also (albeit to a smaller degree) on peripheral physiological levels. Subjective arousal, heart rate, and sweat gland activity increase, whereas skin temperature decreases when addicted individuals are confronted with addiction-relevant cues. Since then, numerous studies have applied cue-reactivity paradigms in functional imaging studies and assessed brain activity during cue exposure (e.g., Chase, Eickhoff, Laird, & Hogarth, 2011; Jasinska, Stein, Kaiser, Naumer, & Yalachkov, 2014; Wilson, Sayette, & Fiez, 2004). The results indicate that the prefrontal cortex, the anterior cingulate cortex, the ventral striatum, the insula, the hippocampus, and the amygdala are involved in cue-reactivity toward substance-related cues. Overall, cue-reactivity paradigms are established research tools for investigating substance-use disorders.

The basis of cue-reactivity is learning mechanisms, in which various cues become associated with the rewarding properties of the drug (Hyman, Malenka, & Nestler, 2006; Tiffany & Wray, 2011). These conditioned cues trigger conditioned emotional/motivational reactions (i.e., cue-reactivity), which provide the basis for experiencing craving and comprise the anticipation of reward or the occurrence of withdrawal symptoms in the case of not consuming the substance. On a neural level, the dopaminergic mesolimbic reward system (especially the ventral striatum) plays a crucial role (Robinson and Berridge 1993, 2008) postulate within their incentive sensitization theory of addiction.

Keywords: cue-reactivity, craving, behavioral addiction, gambling, gaming, buying

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that due to sensitization of the system, cues become more salient and in turn trigger craving reactions. Consumption behavior leads to the desired consequences, which enhances the probability to consume again (operant conditioning) (Drummond, 2001). In the course of addiction, there is a movement from initial ventral striatum activity (representing the hedonic “liking”) to the dorsal striatum (representing the more compulsive and habituated “wanting”) (Everitt & Robbins, 2005). Particularly, the neural responses to cues can predict the risk of relapse after detoxification (Wrase, Grüser, & Heinz, 2006).

The term “addiction” is not restricted to substances. Decades ago, some authors considered different excessive behaviors to be addictive too (von Gebsattel, 1954). To date, only gambling disorder is included as behavioral addiction in the current fifth version of the Diagnostic and Statistical Manual for Mental Disorders (DSM-5) and Internet gaming disorder is included as a research diagnosis (American Psychiatric Association, 2013). However, there is a current debate whether to consider other behaviors, such as Internet-use disorder or buying disorder as addictive conditions as well (Potenza, 2006). Potential similarities between substance-related and behavioral addictions are the urge to perform the behavior, development of tolerance, withdrawal symptoms, unsuccessful attempts to stop or reduce the behavior, and continuation of the behavior despite negative consequences (Grüsser, Poppeleuter, Heinz, Albrecht, & Sass, 2007). This has also been observed for behaviors, such as buying, Internet gaming, and others (Potenza, 2014). To date, research concerning other behaviors than gambling is limited and therefore they are not included in the current official classification systems of mental disorders (Potenza, 2014). In addition, it has been proposed not to pathologize common behaviors (Kardefelt-Winther et al., 2017). However, in recent years, studies were conducted that addressed cue-reactivity in the context of behavioral addictions (e.g., Niu et al., 2016). Persons who perform certain behaviors pathologically are confronted with behavior-relevant cues and their reactions are monitored. For gambling disorder, numerous studies on cue-reactivity exist (e.g., Balodis, Lacadie, & Potenza, 2012; Goudriaan, de Ruiter, van den Brink, Oosterlaan, & Veltman, 2010; McGrath, Dorbeck, & Barret, 2013). The results indicate that individuals with gambling disorder react in a similar way to gambling-relevant cues as individuals with substance-use disorder react to substance-associated cues. More precisely, when confronted with relevant cues, individuals with gambling disorder show subjective reactions (Grüsser, Plötzke, & Albrecht, 2005), increased peripheral physiological arousal (Sharpe, Tarrier, Schötte, & Spence, 1995), and increased brain activity in the dorsolateral prefrontal cortex (Crockford, Goodyear, Edwards, Quickfall, & El-Guebaly, 2005), the striatum (Meng et al., 2014), and the amygdala (Goudriaan et al., 2010). Thus, behavioral addictions might have similarities with substance-use disorders, although no psychotrophic substance is consumed. In a recent meta-analysis, Noori, Cosa Linan, and Spanagel (2016) compared brain activation toward the respective cues for substances, gambling, food, and sex. They found overlapping neural circuits for substance-use disorders and behaviors, i.e., activations in the anterior cingulate gyrus, insula, caudate head, inferior frontal gyrus, middle frontal gyrus, and cerebellum.

The present meta-analysis aims to quantify cue-reactivity for behaviors that are potentially addictive, such as gambling, gaming, buying, Internet usage, exercise, and work. Hypersexual behavior and food addiction were not included because of methodological considerations (see “Methods” section). We hypothesize that people who suffer from those behaviors show higher reactions when they are confronted with behavior-relevant cues compared with healthy control participants. We also hypothesize that people who suffer from those behaviors show higher reactions when they are confronted with behavior-relevant cues compared with control cues.

METHODS

Inclusion criteria

The study procedures were carried out in accordance with the Declaration of Helsinki. We concentrated our literature search on the following behaviors that are considered as potentially addictive: gambling, gaming, Internet usage, buying, exercise, and work. We excluded all types of eating disorders for the following reasons. First, eating includes ingestion and is a natural reward, which is a main difference to the other behavioral addictions. Second, feeding and eating disorders are a separate diagnostic category in the DSM-5. Third, the studies concerning cue-reactivity in eating disorders often include different outcome measures than other studies, for example, the amount of actual consumption. We further excluded studies with potential Internet pornography-use disorder (e.g., Prause, Steele, Stalay, Sabatinelli, & Hajcak, 2015; Voon et al., 2014), because sexuality is also a natural reward and high cue-reactivity toward sexual cues is not restricted to persons with pathological sexual behavior. Furthermore, it has been discussed that pornographic pictures do not represent cues but direct rewards, because seeing them is directly rewarding (Gola, Wordecha, Marchewka, & Sescousse, 2016). In addition, we only included studies that used a classical cue-reactivity paradigm in which patients were exposed to visual or auditory cues, while their subjective, physiological, or neural reactions were measured. That means, we excluded studies in which neuropsychological tasks were applied that were modified with cues (e.g., go/no-go tasks including addiction-relevant cues) or in which the addiction-relevant behavior was actually performed (e.g., playing a video game or simulated gambling).

Literature search

The databases PubMed and PsycInfo were reviewed for suitable articles. The search was conducted using the search terms “gambling addiction,” “pathological gambling,” “gambling disorder,” “gaming addiction,” “pathological gaming,” “gaming disorder,” “buying disorder,” “buying addiction,” “pathological buying,” “compulsive buying,” “shopping addiction,” “pathological shopping,” “shopping disorder,” “compulsive shopping,” “Internet addiction,”

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“pathological Internet use,” “Internet-use disorder,” “Internet gaming disorder,” “Internet gambling disorder,” “Internet-shopping disorder,” “Internet-communication disorder,” “exercise addiction,” “compulsive exercise,” “work addiction,” “compulsive working,” “workaholism,” “cue,” “cue exposure,” “stimuli,” “reactivity,” “craving,” “conditioned withdrawal,” “conditioned responses,” and “urges” in PubMed. In PsychInfo, capacities concerning the number of search terms were limited and the search terms “gaming addiction,” “gaming addiction,” “buying addiction,” “Internet addiction,” “exercise addiction,” “work addiction,” “cue,” “cue exposure,” “stimuli,” “reactivity,” “craving,” “conditioned withdrawal,” “conditioned responses,” and “urges” were used. A search according to Boolean logic was performed, that is, each search term reflecting the behavioral addictions (using the conjunction OR) and each search term reflecting cue-reactivity (using the conjunction OR) were combined using the conjunction AND. Each term was required to be present in the “Title” of the paper. The search period was from January 1995 to September 2017. Further criteria were English language and humans as participants.

The studies were then analyzed as to whether they used the appropriate design, which means that persons with problematic or pathological behavior were investigated with a cue-reactivity paradigm. Cue-reactivity must have been compared between behavior-relevant cues and control cues on a subjective level (e.g., ratings of arousal), peripheral physiological or electroencephal level (e.g., skin conductance responses, heart rate, and electroencephalography), or neural level (e.g., functional magnetic resonance imaging and positron emission tomography). In addition, the references of the selected articles were reviewed for further literature and the papers by relevant authors were screened. Furthermore, an additional judge who was blind to our previous selections screened the references of the selected articles again. We also searched for unpublished data using the ProQuest databases to search for dissertations or master theses (www.proquest.com) and posted a request to send unpublished data on Research Gate (https://www.researchgate.net/home). A total of 18 studies (29 data sets and 510 participants) fulfilled the aforementioned criteria and the relevant data were available and could be included in the meta-analytical calculations. The studies include gambling (eight studies), gaming (nine studies), and buying (one study) disorder. All the studies that were included have been published. The exact searching and selecting procedure is shown in Figure 1. Within the reference list, studies that are included in the meta-analytical calculations are marked with the symbol * (Blanchard, Wulfert, Freidenberg, & Malta, 2000; Crockford et al., 2005; Goudriaan et al., 2010; Han, Hwang, & Renshaw, 2011; Han, Kim, Lee, & Renshaw, 2012; Ko et al., 2009, 2011, 2013; Liu et al., 2017; Lorenz et al., 2012; McGrath et al., 2013; Potenzia et al., 2003; Sharpe et al., 1995; Sodano & Wulfert, 2010; Sun et al., 2012; Thalemann, Wößling, & Grüsser, 2007; Trotzke, Starcke, Pedersen, & Brand, 2014; Wößling et al., 2011).

Coding of variables

The following variables were coded: sample size (N), age, gender, mean, and standard deviation (SD) of cue-reactivity toward behavior-relevant cues and control cues in patients and healthy participants. Few studies used a baseline correction when presenting cue-reactivity data, whereas others did not. We used the raw data of cue-reactivity measures when they were provided and baseline-corrected values only if the raw data were not available. If baseline-corrected data were used, this is indicated in the “Results” section. If more than one type of control cues were used, we used those cues that were more proximal to the addiction-relevant cues [e.g., exciting cues (such as a rollercoaster drive) instead of neutral cues (such as unexciting nature cues)]. Reactions were assessed on a subjective, peripheral physiological, or neural level.

Effect size (ES) calculation and meta-analytical calculations

N, means, and SDs of cue-reactivity were used to calculate Cohen’s d as ES. For the between-design (addiction-relevant cues only; patients vs. control participants), Cohen’s d was calculated as (patients’ mean cue-reactivity toward addiction-relevant cues—control participants’ mean cue-reactivity toward addiction-relevant cues)/pooled SD addiction-relevant cues (DeCoster, 2004; Field & Gillet, 2010). Thus, positive d scores represent higher cue-reactivity toward the addiction-relevant cues in patients than in control participants. For the within-design (patients only; addiction-relevant cues vs. control cues), Cohen’s d was calculated as (mean cue-reactivity toward addiction-relevant cues—mean cue-reactivity toward control cues)/SD cue-reactivity toward control cues (Becker, 1988). Thus, positive d scores represent higher cue-reactivity toward the addiction-relevant cues than toward the control cues. We included the correlation between reactions toward addiction-relevant cues and control cues in the meta-analytical calculations. As not many studies provided these values, we estimated the correlation coefficients from those studies from which they were available and applied them to the rest of the studies. This was done separately for subjective data (two data sets were available) and for peripheral–physiological data (two data sets were available). Random ES model (Hunter & Schmidt, 2004) was used for significance testing, confidence intervals, and χ² tests of homogeneity according to Hedges and Vevea (1998). Significant χ² tests indicate heterogeneity of the results.

For the brain imaging data, a voxel-based meta-analytic approach was used, which is implemented in the software Anisotropic ES-signed differential mapping (ES-SDM), version 5.141 (Radua & Mataix-Cols, 2009; Radua et al., 2012). Two meta-analyses were calculated: (a) one for the comparison between patients and control participants [addicted (addiction-related cues vs. control cues) versus control participants (addiction-related cues vs. control cues)] and (b) other for the contrast addiction-relevant cues against control cues in patients (addiction-related cues vs. control cues). The more conservative contrast against control cues instead of the contrast against baseline was selected in studies, which reported multiple contrasts to avoid biasing meta-analysis results. In all included studies, the same threshold was used throughout the whole brain. Z values from studies, which did not report t values, were transformed using the ES-SDM
web utilities (www.sdmproject.com/utilities). The detailed procedure of ES-SDM software has been described previously (Radua & Mataix-Cols, 2009; Radua et al., 2012). The program converts the peak \( t \) values to Hedges ES and uses those ESs and the peak coordinates to recreate maps of the ES of the BOLD response for the contrast of interest. This algorithm applies an anisotropic non-normalized Gaussian kernel to the voxels close to the peak. Afterward, a voxel-wise random effects meta-analysis is conducted, which weights the studies for sample size and variance. Default ES-SDM kernel size and thresholds were used as recommended by Radua et al. (2012) (FWHM =20 mm, \( p < .005 \), peak height \( Z > 1, k > 10 \) voxels). For purposes of visualization, results were superimposed on a single-subject Talairach template (colin27) (Kochunov et al., 2002), using the Multi-image Analysis GUI (version 3.8; Mango, Research Imaging Institute, University of Texas Health Science Center at San Antonio, TX, USA).

**Publication bias**

As a first rough indicator of publication bias, the fail-safe \( N \) procedure by Rosenthal (1979) was used. This procedure estimates the number of unpublished studies needed to turn a significant ES into a non-significant one. Then, Begg and Mazumdar’s (1994) rank correlation test was applied. This test quantifies the relationship between ES and sample size, and a significant result indicates potential publication bias. Funnel plots as proposed, for example, by Vevea and Woods (2005) were not created because of the small number of data sets that could be included into each category (Higgins & Green, 2011).

Figure 1. Flow chart of the literature search. *This was the case in one study in which the only dependent variable was the subjective urge to perform the behavior. Subjective urges elicited by cues were excluded in the current analysis because the comparison between addiction-relevant cues and addiction-irrelevant cues is inappropriate in this case. Addiction-irrelevant cues should not elicit an urge to perform the addictive behavior and this would lead to an overestimation of effect size.*
RESULTS

The results of the meta-analytical calculations are presented according to the respective cue-reactivity measures. For subjective and peripheral physiological and electroencephal measures, a positive $d$ score means that patients react stronger toward addiction-relevant cues than control participants and that patients react stronger toward addiction-relevant cues than to control cues. In contrast to the meta-analysis by Carter and Tiffany (1999), studies on functional magnetic resonance imaging are also included here. Furthermore, there is a greater variety in peripheral physiological and electroencephal recordings. Therefore, we sorted our cue-reactivity measures according to the following three outcome measures: (a) subjective measures, (b) peripheral physiological and electroencephal measures, and (c) functional brain imaging data. After providing the pooled effects of gambling, gaming, and buying, we provide descriptive results for each of the addictions separately. We then report effects after excluding data on buying disorder, because only a single study on buying disorder was available. Finally, we report effects excluding data of the two studies that used baseline-corrected values instead of raw values.

Characteristics of each study

In Table 1, we provide the main characteristics of each study.

Subjective cue-reactivity

Variables that assessed how strong the specific cues elicited arousal, excitement, etc., were included, whereas variables that assessed how strong the cues elicited the urge to perform the behavior were excluded. Control cues are not suited to elicit an urge to perform a behavior. For example, the picture of a buffalo should not elicit the urge to buy the buffalo in individuals with buying disorder; the video of a landscape should not elicit the urge to gamble in individuals with gambling disorder. The inclusion of questions such as “how strong does the cue elicit the urge to perform a certain behavior?” would lead to overestimated effects. Furthermore, average scores were built in case that more than one subjective measure was assessed. If a questionnaire with different items, such as “pleasant,” “stimulated,” “satisfied,” and “high,” was presented after cue presentation, we averaged the scores of these items and only calculated a single ES $d$.

Addiction-relevant cues in patients versus control participants. For the six data sets available, an ES of $d = 0.84$ was observed ($k = 6$, lower $d = 0.19$, upper $d = 1.49$, standard error $= 0.33$, $z = 2.54$, $p = .01$). The test for homogeneity was not significant ($Q = 3.80$, $df = 5$, $p = .58$). Rosenthal’s fail-safe was $N = 76$. Begg and Mazumdar’s rank correlation test was not significant ($r = 0.07$). Out of the six data sets, four included data on gambling, and an ES of $d = 0.45$ was observed; the ES of the single study on gambling was $d = 1.00$; the ES of the single study on buying was $d = 1.96$. Excluding data on buying disorder results in an ES of $d = 0.52$ (remains significant). Excluding the two studies that used baseline-corrected values results in an ES of $d = 1.07$ (remains significant).

Addiction-relevant cues versus control cues in patients. For the seven data sets available, an ES of $d = 0.39$ was observed ($k = 7$, lower $d = -0.09$, upper $d = 0.86$, standard error $= 0.24$, $z = 1.59$, $p = .11$). The test for homogeneity was not significant ($Q = 6.69$, $df = 6$, $p = .35$). Rosenthal’s fail-safe was $N = 16$. Begg and Mazumdar’s rank correlation test was not significant ($r = 0.14$, $p = .65$). Out of the seven data sets, five included data on gambling, and an ES of $d = 0.09$ was observed; the ES of the single study on gaming was $d = 1.74$; the ES of the single study on buying was $d = 0.87$. Excluding data on buying disorder results in an ES of $d = 0.29$ (remains insignificant). Excluding the two studies that used baseline-corrected values results in an ES of $d = 0.61$ (reaches significance).

Peripheral physiological and electroencephal cue-reactivity

As peripheral physiological measures, heart rate, skin conductance level, skin conductance responses, skin resistance level, systolic blood pressure, and electromyography were used. In addition, electroencephalography was included.

Addiction-relevant cues in patients versus control participants. For the seven data sets available, an ES of $d = 0.61$ was observed ($k = 7$, lower $d = 0.16$, upper $d = 1.06$, standard error $= 0.23$, $z = 2.65$, $p < .01$). The test for homogeneity was not significant ($Q = 6.12$, $df = 6$, $p = .41$). Rosenthal’s fail-safe was $N = 51$. Begg and Mazumdar’s rank correlation test was not significant ($r = 0.42$, $p = .18$). Out of the seven data sets, five included data on gambling (two studies assessed heart rate, one assessed skin conductance level, electromyography, and electroencephalography respectively), and an ES of $d = 0.60$ was observed; the ES of the single study on gaming (electroencephalography) was $d = 0.76$; the ES of the single study on buying (skin conductance responses) was $d = 0.63$. Excluding data on buying disorder results in an ES of $d = 0.62$ (remains significant). Excluding the two studies that used baseline-corrected values results in an ES of $d = 0.92$ (remains significant).

Addiction-relevant cues versus control cues in patients. For the 11 data sets available, an ES of $d = 0.47$ was observed ($k = 11$, lower $d = 0.00$, upper $d = 0.94$, standard error $= 0.24$, $z = 1.96$, $p = .05$). The test for homogeneity was not significant ($Q = 10.11$, $df = 10$, $p = .43$). Rosenthal’s fail-safe was $N = 59$. Begg and Mazumdar’s rank correlation test was not significant ($r = -0.02$, $p = .94$). Out of the 11 data sets, nine included data on gambling (four studies assessed heart rate, one assessed systolic blood pressure, skin resistance level, skin conductance level, electromyography, and electroencephalography, respectively), and an ES of $d = 0.36$ was observed; the ES of the single study on gaming (electroencephalography) was $d = 1.86$; the ES of the single study on buying (skin conductance responses) was $d = 0.07$. Excluding data on buying disorder results in an ES of $d = 0.51$ (changes from significant to marginally significant). Excluding the two studies that used baseline-corrected values results in an ES of $d = 0.65$ (remains significant).

Cue-reactivity in functional brain imaging studies

All studies that were included used functional magnetic resonance imaging. All 3D coordinates of activation
maxima were reported in standardized stereotactic space (Montreal Neurological Institute or Talairach).

**Addiction-relevant cues in patients versus control participants.** For the 10 data sets available, contrasting activations in patients and control participants, the analysis revealed six clusters, shown in Figure 2 and described in Table 2. In the right hemisphere, clusters were found in the median cingulate cortex (anterior cingulate, posterior cingulate, and paracingulate gyri), inferior frontal gyrus, and caudate nucleus. In the left hemisphere, activations were found in precentral gyrus and caudate nucleus. Moreover, activation in the subgenual cingulate was found. Out of the 10 data sets, two included data on gambling. With a threshold probability of \( p < .01 \), the analysis revealed significant activations in right striatum, right fusiform gyrus, median cingulate gyrus, middle occipital gyrus, and right inferior frontal gyrus. Eight studies included data on gaming. Significant clusters (with \( p < .01 \)) were found in median cingulate, right inferior frontal gyrus, precentral gyrus, right caudate nucleus, left precentral gyrus, right superior temporal gyrus, corpus callosum, and right precuneus.

**Addiction-relevant cues versus control cues in patients.** For the six data sets available, the ES-SDM meta-analysis revealed six clusters of significant convergence across studies when contrasting addiction-relevant cues with...
control cues, which are shown in Figure 2 and described in Table 3. In the right hemisphere, a cluster with activations in the inferior frontal gyrus, caudate nucleus, angular gyrus, and precuneus were found and in the left hemisphere, a cluster with activations in inferior network (inferior longitudinal fasciculus, calcarine fissure, and fusiform gyrus) and inferior frontal gyrus were found. Out of the six data sets, one included data on gambling. In this study (Crockford et al., 2005), significant activations were found in bilateral precuneus, inferior parietal lobule, occipital cortex, cuneus, lingual gyrus, fusiform gyrus, right inferior frontal gyrus, and right parahippocampal gyrus. Five studies included data on gaming. The SDM meta-analysis revealed significant clusters (with p < .01) in bilateral inferior frontal gyrus, corpus callosum, right caudate nucleus, right angular gyrus, and bilateral cingulate gyrus.

**DISCUSSION AND CONCLUSIONS**

**Summary of evidence and interpretation**

Individuals with gambling, gaming, and buying disorders show higher cue-reactivity toward addiction-relevant cues compared with healthy control participants. Pooled data for subjective ratings of cue-reactivity reached significance. The ES was large. Pooled data for peripheral physiological and electroencephal measures also reached significance. The ES was medium. On a neural level, significant activations were observed in the caudate nucleus, the median cingulate cortex, the subgenual cingulate, the inferior frontal gyrus, and the precentral gyrus. Individuals with gambling, gaming, and buying disorders also show higher cue-reactivity toward addiction-relevant cues compared with control cues.
except for subjective ratings. Pooled data for subjective ratings of cue-reactivity did not reach significance. The ES was only small to medium. Pooled data for peripheral physiological and electroencephal measures reached significance. The ES was medium. On the brain level, significant activations were observed in the caudate nucleus, inferior frontal gyrus, angular gyrus, precuneus, and inferior network. Overall, the results indicate that cue-reactivity is not only restricted to substance-use disorders but also occurs in behavioral addictions. The only non-significant finding was the subjective rating of addiction-relevant cues compared with control cues within the patient group. The results are — overall — in line with meta-analyses on cue-reactivity in substance-use disorders (Carter & Tiffany, 1999; Noori et al., 2016).

It has been proposed that similar mechanisms underlie the development and maintenance of substance-use disorder and behavioral addictions. As outlined in the introduction, conditioning processes, such as classical and operant conditioning, can lead to desires of substances as well as behaviors (Grüsser et al., 2007). Relevant cues are associated with the consequences of a behavior (by classical conditioning) and elicit a motivational state (cue-reactivity) that leads to the desire to perform the behavior (craving). The actual performance of the behavior leads to the desired consequences, which enhances the probability to perform it again (operant conditioning) (Drummond, 2001). The results of the functional imaging studies indicate increased activation of regions that are associated with the anticipation and processing of rewards, and reward-related decision-making, such as the caudate nucleus and the median cingulate (Liu, Hairston, Schrier, & Fan, 2011). The caudate nucleus is part of the dorsal striatum and this region is associated with cue-reactivity processing in chronic-addicted individuals who show habituated, compulsive consumption patterns. It is well described in addiction research that there is a movement from initial ventral striatum activity to the dorsal striatum in the course of addictions (Everitt & Robbins, 2005). Not all behavioral addictions are officially classified and therapeutic interventions are limited, several patients may suffer from chronic addiction. Furthermore, the inferior frontal gyrus is involved in the detection of important cues (Hampshire, Chamberlain, Monti, Duncan, & Owen, 2010). Current results emphasize that the increased salience for addiction-relevant cues, which was first established in the studies of substance-use disorder, can also be assumed for behavioral addictions.

Recently, gambling disorder has been moved from impulse control disorders to the category of substance-related and addictive disorders in the DSM-5 and Internet gaming disorder has been listed as condition that needs further research and potentially belongs to this new category. Other behaviors, such as buying disorder, have not been classified to date, because the empirical evidence concerning these behaviors is still limited (Potenza, 2014). However, meanwhile, some new studies have been published and cue-reactivity paradigms have been established in persons with problematic behaviors other than gambling. In the current meta-analysis, ESs that were observed in patients with gambling and buying disorders were even higher than those that were observed in patients with gambling disorder in most of the analyses. This is a descriptive finding, but it is important to note that cue-reactivity in gaming and buying disorders might even exceed cue-reactivity in the only disorder that is officially classified as a behavioral addiction in the DSM-5. On the brain level, similar regions are involved in cue-reactivity in patients with gambling disorder and those with gaming disorder, namely right inferior frontal gyrus, striatum (caudate nucleus), and cingulate cortex. The results indicate that similar mechanisms may be involved in the development and maintenance of gambling, gaming, and buying disorders. Consequently, gambling and buying disorders should also be considered as candidates for the category substance-related and addictive disorders in future versions of the DSM and the ICD.

Several researchers emphasize that common behaviors should not be pathologized and that not all excessive behaviors are addictive in nature (Kardefelt-Winther et al., 2017). For example, behavior, that is potentially harmful but results from a willful choice, should not be regarded as addictive. Future work should attempt to differentiate cue-reactivity in persons who are addicted to a behavior from those who are engaged in a certain behavior as the results of a willful choice.

Limitations

A relevant limitation of the meta-analysis is that the overall number of studies is small. The possibility exists that there are studies that were not covered by any of our database searches or that are published nowhere. Furthermore, in some studies, relevant data were not provided in the form required, which also led to a reduction of data available. We did not find appropriate studies for all potential behavioral addictions we intended to include. Thus, we did not find any studies that used cue-reactivity paradigms in clinical samples with potential further behavioral addictions, such as exercise addiction or workaholism. In addition, the number of studies that investigated cue-reactivity in each category was small. Therefore, potential differences between gambling, gaming, and buying disorders were only reported on a descriptive level and on the brain level with an increased threshold probability of $p < .01$. Only one of the studies reported data on buying disorder. Therefore, we performed a sensitivity analysis using the leave-one-out method and excluded data from this study. The results basically remained stable.

Many of the studies included contain more than one data set (e.g., subjective and physiological data), which means that the overall number of data sets is larger than the number of studies. Including multiple data sets of the same study has also been regarded as a problem. However, Ishak, Platt, Joseph, and Hanley (2008) assumed that dependencies do not change the results and so we did not change the common meta-analytical calculation methods. The Begg and Mazumdar’s (1994) rank correlation test did not indicate publication bias. However, in small meta-analyses, a non-significant result does not guarantee the absence of publication biases. Funnel plots as proposed, for example, by Vevea and Woods (2005) are principally suited for small meta-analyses, but at least 10 data sets should be included.

in each calculation (Higgins & Green, 2011), which was not the case for all calculations.

Another major limitation of the current meta-analysis is the heterogeneity of studies. Some studies were conducted with pictures as cues, others used videos or audiotapes. The dependent variables also varied among studies. For example, the subjective cue ratings represented experiences of arousal elicited by the cue, liking the cue, or the urge to perform the behavior when exposed to the cue. Variables that assessed the urge to perform the behavior had to be excluded from the analysis, because control cues were unsuited for this assessment. Therefore, no conclusions about subjective urges can be drawn here. Furthermore, subjective ratings were frequently assessed by questionnaires, which contained more than one question and were averaged for the analysis. Therefore, the overall value for subjective responses is unspecific. The same limitation occurred for the physiological data, which were combined and included various outcome measures (skin conductance, heart rate, blood pressure, electromyography, and electroencephalography). The control stimuli also differed between studies, i.e., some used neutral cues as control cues, whereas others used emotional addiction-irrelevant cues as control cues. Another shortcoming is that the studies used different contrasting procedures. While for the majority of studies, the raw data were used for the meta-analysis, two of the studies only reported results that were baseline-corrected. Including them led to decreased effects, so our calculations were conservative in this regard. Furthermore, most studies contrasted patients’ cue-reactivity to addiction-relevant cues versus control cues and also contrasted patients’ cue-reactivity to addiction-relevant cues with a healthy control group. However, some studies only reported either of the above category. The studies also differed concerning the order of stimulus presentation (e.g., same for all participants vs. randomized). Concerning the participants, the proportion of female participants was low except for individuals with buying disorder. Furthermore, the inclusion of participants differed, i.e., while some studies investigated only patients with the same specific problematic behavior (e.g., who play a specific online game) and exposed them to their preferred cues, other studies investigated patients with mixed behaviors (e.g., who play any online game). Accordingly, some studies used cues that strongly resembled the patients’ preferred cues, whereas in other studies, cues were less specific. A limitation that directly results from the fact that the behavioral addictions (besides gambling disorder) are not included in the official diagnostic systems is that not all patients had an official diagnosis. We included only studies in which the participants showed at least problematic behavior, but the threshold for defining a problematic behavior may have varied among the studies.

Implications for future research

More knowledge concerning cue-reactivity in behavioral addictions would allow the application of neuropsychological tasks, which are modified with behavior-relevant cues. In patient groups with behavioral addictions, relevant cues should influence cognitive processes (Brand, Young, Laier, Wölfling, & Potenza, 2016). The assumption that cue-reactivity triggers craving reactions, which can lead to poor executive functions (such as reduced inhibitory processes) and finally lead to poor decision-making (performing the behavior, which is short-term rewarding, but has negative long-term consequences), could be tested in pathologies other than gambling and gaming. Furthermore, it is rational to study decision-making processes directly when confronted with addiction-relevant cues. Some recent studies in this field exist for buying (Raab, Elger, Neuner, & Weber, 2011) and gambling disorders (e.g., Miedl, Büchel, & Peters, 2014).

Importantly, research on cue-reactivity in behavioral addictions could contribute to the debate whether behavioral addictions other than gambling exist and what constitutes a behavioral addiction. To avoid an inflationary use of the term “behavioral addiction,” it is essential to differentiate them from behaviors that are executed willfully, such as hobbies or professions (Kardefelt-Winther et al., 2017). Therefore, it would be useful to compare reactions to cues between persons whose excessive behavior is part of a willful choice (e.g., professional poker players and professional athletes) and persons whose excessive behavior cannot be explained by a willful choice (e.g., patients with gambling disorder who lose all their money and neglect their jobs and families and patients with running addiction who neglect their jobs, families, and health). The studies would profit from the assessment of cue-reactivity combining physiological and neural responses with fine-grained subjective measures. Subjective measures should be able to differentiate the hedonic “liking,” and the more compulsive “wanting” component toward cues, which could be related to the specific physiological or neural responses. Such research could contribute to the debate on whether each hedonic behavior can become addictive and what discriminates excessive (but functional) from addictive (dysfunctional) behavior.

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References marked with an asterisk were included in the meta-analytical calculations.


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