

Let's Open the “Decision-Making” Umbrella: A Framework for Conceptualizing and Assessing Features of Impaired Decision Making in Addiction

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1 **Abstract**

2 Decision-making impairments play a pivotal role in the emergence and maintenance of
3 addictive disorders. However, a sound conceptualization of decision making as an umbrella
4 construct, encompassing its cognitive, affective, motivational, and physiological
5 subcomponents, is still lacking, preventing an efficient evaluation of the heterogeneity of
6 decision-making impairments and the development of tailored treatment. This paper thus
7 unfolds the various processes involved in decision making by adopting a critical approach of
8 prominent dual- or triadic-process models, which postulate that decision making is influenced
9 by the interplay of impulsive-automatic, reflective-controlled, and interoceptive processes.
10 Our approach also focuses on social cognition processes, which play a pivotal role in decision
11 making and addictive disorders but were largely ignored in previous dual- or triadic-process
12 models. We propose here a theoretical framework in which a range of coordinated processes
13 are first identified on the basis of their theoretical and clinical relevance. Each selected
14 process is then defined before reviewing available results underlining its role in addictive
15 disorders (i.e., substance use, gambling, and gaming disorders). Laboratory tasks for
16 measuring each process are also proposed, initiating a preliminary process-based decision-
17 making assessment battery. This original approach may offer an especially informative view
18 of the constitutive features of decision-making impairments in addiction. As prior research
19 has implicated these features as risk factors for the development and maintenance of addictive
20 disorders, our processual approach sets the scene for novel and transdiagnostic experimental
21 and applied research avenues.

22 **Keywords:** decision making; addiction; dual-process model; assessment battery;
23 transdiagnostic and processual approach; social cognition.

1 **Introduction**

2 Being regularly confronted with temptations, human beings are constantly forced to make
3 choices, some bad and others good. In view of the importance of decision making, various
4 mechanisms or processes underlying it have been studied for many years from different
5 disciplinary perspectives (e.g., philosophy, psychology, economy, neurosciences, artificial
6 intelligence). Crucially, impairment in decision making has long been established as a
7 hallmark of addictive disorders (Bechara 2005; Bickel et al. 2018; Goudriaan et al. 2005),
8 which are commonly understood as pathologies of “decision-making” (Redish et al. 2008),
9 “choices” (Kalivas and Volkow 2005), or “temporal horizon” (Bickel et al. 2006).

10 Decision making is often viewed as the result of an active and conscious deliberative
11 process in which individuals balance the short-term and long-term benefits of a choice in light
12 of the information available (Evans 2003; Figner et al. 2009). In such an approach, decision
13 making is considered an active, voluntary, and cognitively demanding process. Yet, the
14 cognitive abilities of human beings are naturally limited, and the knowledge available to
15 guide an informed decision largely varies by individual or environmental factors. In
16 accordance with this, economist Herbert Simon theorized the influential limited rationality
17 theory (1979; see also Gigerenzer and Goldstein 1996), stipulating that individuals are
18 rational agents who generally favor the first satisfying option rather than engaging in a full,
19 comprehensive, and cognitively demanding deliberative process. Thus, adaptive decision
20 making not only relies on cognitive (and contextual factors), but also depends on the
21 influence of affective and physiological processes. Important corroboration of this view is
22 provided by the somatic marker hypothesis (Damasio 1996). According to Damasio, decision
23 making relies on a cost-benefit analysis related to a given action, but is also influenced by
24 “somatic markers,” i.e., anticipatory physiological and emotional reactions generated on the
25 basis of the outcomes associated with similar past choices. These somatic markers are

1 especially supposed to guide choices in uncertain contexts, where no detailed cost-benefit
 2 analysis is possible. Beyond the involvement of cognitive, affective, and physiological
 3 mechanisms, decision making is generally considered to depend on the mutual and conflicting
 4 actions of automatic (unconscious) and controlled (voluntary) processes (Epstein 1994; Evans
 5 2003; Strack and Deutsch 2004). The following section addresses the key distinction between
 6 “automatic” versus “controlled” determinants of behavior with regard to addictive disorders
 7 as assumed in the so-called dual-process models of decision-making (e.g., Bechara 2005;
 8 Mukherjee 2010; Schiebener and Brand 2015; Sloman 1996).

10 **Dual-Process Models of Addiction: Origins, Evolution, and Limitations**

11 The dual-process approach has become massively prominent in the field of addictive
 12 behaviors, and almost all influential models consider the dichotomy between impulsive-
 13 automatic and reflective-controlled determinants of behavior (e.g., Koob and Le Moal 2001;
 14 Volkow et al. 2011). In fact, specific dual-process models have been explicitly developed to
 15 account for addictive disorders (e.g., Bechara 2005; Evans and Coventry 2006; Fleming and
 16 Bartholow 2014; Noël, Brevers et al. 2013a, 2013b). According to the prominent Bechara’s
 17 model (2005), addiction results from the product of an imbalance between two distinct but
 18 interacting cognitive and cerebral systems controlling decision making: an “impulsive”
 19 system related to amygdala/striatum activity and driven toward immediate positive
 20 reinforcement (e.g., getting high) or negative reinforcement (e.g., reducing withdrawal), and a
 21 “reflective” system related to prefrontal activity and focused on future prospects (e.g., the cost
 22 and benefits of remaining sober). Such models typically postulate that in addictive disorders,
 23 the impulsive-automatic system tends to be overactivated (mainly in reaction to internal or
 24 external cues), whereas the mechanisms related to the optimal functioning of the reflective-
 25 controlled system are impaired (e.g., inhibitory control deficit). More recently, dual-process

approaches also inspired the development of models accounting for conditions such as gambling and gaming disorders (Brand et al. 2016; Brevers and Noël 2013; Dong and Potenza 2014; Wei et al. 2017), in which the interplay between automatic and reflective processes is, like that in substance-related addictions, postulated to play a pivotal role in the onset and perpetuation of the condition. According to several authors (e.g., McClure and Bickel 2014; Schiebener and Brand 2017), disentangling the various processes involved in impulsive-automatic versus reflective-controlled systems is an important step toward better understanding and treatment of addictive disorders.

In recent years, to better account for specific features of addiction such as craving or urges (Verdejo-García et al. 2012), Noël, Brevers et al. (2013a, 2013b) extended the dual-process perspective to a “triadic” model integrating a third system related to interoceptive processes. The triadic model exemplifies how dual-process models are currently evolving based on the type of behavior they aim to explain (here addictive disorders). This interoceptive system, which mainly depends on the activity of the insular cortex (in particular its anterior part), promotes addictive behaviors by translating interoceptive signals into subjective experiences, such as feelings of desire, anticipation, urge, or craving (e.g., Naqvi et al. 2007; Verdejo-García et al. 2012). According to Noël, Brevers et al. (2013a, 2013b), such interoceptive information is susceptible to fostering activity of the impulsive-automatic system while undermining the efficacy of the reflective-controlled system, promoting dysregulated and compulsive drug seeking. Furthermore, it has also been suggested that dysfunction in the interoceptive system engenders poorer self-awareness, which can reinforce denial, reduced perceived need of treatment, or continuous substance abuse despite deleterious health consequences (Goldstein et al. 2009; Goldstein and Volkow 2011). Although this triadic model constitutes an inspiring evolution of previous dual-process models and sophisticates their application to addictive disorders, it still lacks empirical

support to confirm the role played by the insular system and its moderating function between the limbic and prefrontal areas.

Nevertheless, despite the popularity of these models, experts in the field of decision-making research have recently initiated a dismissive appraisal of the dual-process approach's prospects as a viable and useful model of decision making (e.g., Hommel and Wiers 2017; Keren 2013; Keren and Schul 2009; Newell and Shanks 2014). One serious criticism is the lack of well-defined constructs that renders impossible the generation of testable predictions within the dual-process framework (for extended discussion, see Keren 2013; Keren and Schul 2009). Moreover, the fuzzy conceptual boundaries between the two systems also seemingly challenge the claim that these two systems are extricable (Hommel and Wiers 2017; Keren 2013; Keren and Schul 2009). Likewise, recent research has suggested that most of the processes that are assumed to fall under the label of one of two systems—at least, according to the dual-process perspective—often involve features of both systems. For instance, although the dual-process perspective conceptualizes attentional biases as a core feature of the “automatic-impulsive system” (e.g., Wiers, Bartholow et al. 2007; Wiers et al. 2013), the modifications of the very central features of the so-called reflective system, such as via a transient increase in cortical activity within the prefrontal cortex, modulate attentional biases (Heeren et al. 2017). Such a criticism of the dual-process perspective should not come as a surprise; on the contrary, it is clearly in line with recent empirical and theoretical work in cognitive science defying the arbitrary distinction between controlled and automatic processes and instead promoting a continuum within a unique functionally integrated cognitive system (Pessoa 2017). Accordingly, because it has become almost impossible to categorize any processes or tasks as specifically pertaining to one the two systems, recent models of decision making have evolved toward a unitary model of action control, driven by a continuum

between intention and automaticity, working as a unique functionally integrated decision-making system (e.g., Hommel and Wiers 2017).

In addition to the aforementioned limitations, by focusing on the imbalance between reflective and automatic systems to explain addictive behaviors, dual-process models, as well as their extension in the triadic model, have neglected other key factors involved in the development and maintenance of addictive disorders, among which are deficits in social cognition-related processes. This issue is particularly unfortunate because the experimental data available have reported that emotional and interpersonal impairments are widely present among patients presenting addictive disorders. In addition, these socio-cognitive impairments have detrimental consequences on various treatment outcomes such as relapse rate (Gordon and Zrull 1991; Zywiak et al. 2003). Moreover, beyond addiction, many studies have pointed out the critical role of social cognition in decision making (for a review, see Frith and Singer 2008). However, despite recent blooming of these social cognition subcomponents in the exploration of social cognition in psychiatric disorders such as schizophrenia (e.g., Green et al. 2015) and autism (e.g., Leekam and Ramsden 2006), their exploration is still in its infancy in addictive disorders.

Reframing Features of Impaired Decision Making: Implications for Addictive Disorders

Addiction is at its core a problem of decision making. Indeed, a large number of case-control studies highlighted that individuals with substance use and addictive disorders (e.g., gambling disorder, gaming disorder) also showed impairment in laboratory measures of decision making (Biernacki et al. 2016; Brevers, Bechara et al. 2013; Goudriaan et al. 2004; Schiebener and Brand 2017; Yücel and Lubman 2007). Existing research also emphasized that more pronounced decision-making impairments are linked with more severe addictive disorders (Bechara et al. 2001; Bickel et al. 2014; Krmpotich et al. 2015) and that decision-

1 making performance allows discrimination between individuals exhibiting recreational and
2 those presenting pathological patterns of cocaine use or gambling involvement (Grant et al.
3 2011; Hulka et al. 2014). A corpus of data also supports the prognosis value of laboratory
4 measures of decision making for various clinical outcomes, such as poorer treatment
5 compliance or increased relapse rates (Álvarez-Moya et al. 2011; Bowden-Jones et al. 2005;
6 Domínguez-Salas et al. 2016; Stevens et al. 2014). Taken together, these data call for the
7 consideration of decision-making-related processes in the diagnosis, treatment, and
8 prevention of addictive disorders (Yücel and Lubman 2007).

9 Yet, a serious limitation of existing studies is that they often failed to conceptualize
10 decision making as an umbrella construct involving a wide range of distinct cognitive,
11 affective, motivational, and physiological processes. The upshot was that they did not often
12 lead to tailored interventions that targeted specific impaired mechanisms (e.g., neurocognitive
13 rehabilitation). Indeed, the classic laboratory tasks generally used to identify decision-making
14 deficits [e.g., Iowa Gambling Task (Bechara et al., 1994), Cambridge Gambling Task (Rogers
15 et al., 1999), Game of Dice Task (Brand et al., 2005), Balloon Analogue Risk Task (Lejuez et
16 al., 2002), delay reward discounting task such as the Monetary Choice Questionnaire (Kirby
17 et al. 1999), Information Sampling Task (Clark et al. 2006)] are “generic” and
18 multidetermined, meaning that performances on these tasks are influenced by many lower
19 order mechanisms, such as inhibitory control, working memory components (e.g., central
20 executive), ability to project oneself in the future, sensitivity to reward, or emotion-laden
21 impulsivity (e.g., Billieux et al. 2010; Cui et al. 2015; Noël et al. 2007). Consequently, tasks
22 that are considered measures of outcomes of poor decision making provide unspecific
23 information about the nature of decision-making-related deficits presented by patients, which
24 brings into question their clinical usefulness.

Given the aforementioned literature, our objective is to unfold the various processes underlying decision making, capitalizing on recent research in addiction that goes beyond the old-fashioned dual-process perspective by including features of the so-called impulsive-automatic and reflective-controlled processes from this dual-process perspective, as well as interoceptive and social cognition processes. We believe that our approach will also likely contribute to a better understanding of the specific processes involved in the cognitive steps involved in decision making (Verdejo-García et al. 2018), such as (a) the formation of preferences (e.g., developing a preference for a specific option by mentally computing and comparing the subjective value of each available option), (b) choice implementation (e.g., allocating a response to the preferred option), and (c) feedback processing and learning (e.g., shaping behavior and subsequent choices according to the outcomes of previous experiences. Our rationale for the selection of the processes was twofold. First, our decision was based on the relevance of these processes with regard to previous prominent frameworks of decision making in addiction. Second, we also relied on the robustness of the evidence linking these processes to addictive disorders, that is, the processes involved via evidence in the diagnosis, treatment, and prevention of addictive disorders. We considered as eligible all data obtained in substance use disorders and gambling and gaming disorders (American Psychiatric Association 2013; World Health Organization 2018). We first define each selected process before reviewing the available results underlining the role played by this process in addictive disorders, and then we describe a valid task that measures this process, leading to a preliminary comprehensive and process-based decision-making assessment battery.

Proposal for a taxonomy of transdiagnostic features of decision making in addiction and a related assessment battery

Reward Sensitivity: Definition and Role in Addictive Disorders The reward system is crucial for processing the salience of incoming stimuli and regulating motivational, emotional, and decision-making abilities (Volkow et al. 2010). Efficient reward processing is thus indispensable in healthy humans to initiate and regulate appetitive behaviors, and the proposal that addictive disorders are closely related to changes in reward sensitivity mechanisms is at the heart of the neurobiological models of addiction (Koob and Le Moal 2001; Robinson and Berridge 2003; Volkow et al. 2003). As reward sensitivity is centrally related to limbic structures and associative learning, it is considered part of the impulsive-automatic system from the perspective of the dual-model process, being related to appetitive associations that cannot be directly controlled by the reflective system. However, recent models support the view that reward processing in fact plays a pivotal role in both automatic and reflective aspects of behavior (see O'Doherty et al. 2017), as it is at the very core of any type of goal-directed behavior. Other psychological models have linked reward sensitivity with impulsivity (Whiteside and Lynam 2001) or sensation-seeking traits (Blum et al. 1996), despite these constructs being theoretically and empirically distinct (e.g., sensation seekers often characterized as having increased impulsivity toward specific addiction-related stimuli without presenting globally increased reward sensitivity). Reward sensitivity is thus a key concept in addictive disorders, which should be taken into account and validly measured, notably in view of its link with impulsivity and decision making.

While reward sensitivity toward drug-related stimuli has been extensively explored by using attentional biases and implicit attitude paradigms (see related sections), the more general sensitivity to reward and punishment has been far less investigated in addictive disorders. Several tasks have been developed to offer a specific measure of reward/punishment sensitivity [e.g., Learning Reinforcement Task (Frank et al. 2004); see below), but they have not yet been used in addictive disorders. The only currently available

data regarding reward sensitivity have been obtained through multidetermined tasks, in particular the Balloon Analogue Risk Task (Lejuez et al. 2002). Results obtained with this task should be taken with caution, as performance on this task can be influenced by factors unrelated to reward sensitivity (e.g., global executive functioning). Moreover, reward sensitivity is merged with risk-taking tendency in this task, despite these two concepts not being considered isomorphic. Nevertheless, increased reward sensitivity (i.e., increased risk taking to obtain higher monetary reward) has been documented to significantly predict heavy alcohol use in young individuals (Fernie et al. 2010). Conversely, long-term alcohol-dependent patients present reduced risk taking in such tasks (Ashenhurst et al. 2011; Campbell et al. 2013), thus suggesting reduced reward sensitivity toward non-alcohol-related rewards. Similar results were found in nicotine-dependence studies, where young adolescent smokers showed increased reward sensitivity (Schepis et al. 2011), whereas adult smokers presented reduced reward-seeking strategies (Lejuez et al. 2005; Ryan et al. 2013). Finally, a recent study showed that individuals characterized by a proneness toward excessive video gaming presented heightened risk taking on the Balloon Analogue Risk Task (Weinstein et al. 2016). While promising, these results have not yet been extended to other addictive disorders, where the exploration of reward sensitivity remains scarce. For example, MDMA users (Hopko et al. 2006) and adolescent gamblers (Cosenza et al. 2017) present increased reward sensitivity; conversely, current opioid (Paydary et al. 2016), cannabis (Gonzalez et al. 2012), or cocaine (Bornovalova et al. 2005) use does not lead to significant reward sensitivity changes. Although the knowledge about substance-related reward sensitivity is thus largely developed, more global reward sensitivity remains to be explored with more specific tasks in addictive disorders.

Measuring Reward Sensitivity A relevant task to measure reward sensitivity is the Learning Reinforcement Task (Frank et al., 2004), which starts with a training phase in which symbols

are repeatedly presented in three pairs (A-B; C-D; E-F). For each trial, participants have to choose one of the two symbols, which leads to feedback (“correct”/“incorrect”). While initially choosing randomly, participants progressively learn that each symbol is associated with positive feedback probability (80%-20% for A-B symbols, 70%-30% for C-D, and 60%-40% for E-F). A testing phase then starts with the same symbols but with reorganized pairs (e.g., A-C; B-D); the frequency of A versus B symbol selection shows whether the participant has mostly learned by using negative or positive feedback. Indeed, participants who learn through positive feedback will mostly choose the “A” symbol (related to most positive feedback in the learning phase), indicating high reward sensitivity, whereas participants who learn through negative feedback will mostly avoid choosing the “B” symbol (related to most negative feedback in the learning phase), indicating high punishment sensitivity.

Attentional Biases: Definition and Role in Addictive Disorders Attentional bias can be defined as preferential attentional allocation toward a specific type of stimuli. Arguably, most influential models of addiction have proposed that, through associative learning processes, addiction-related cues acquire strong incentive salience that, in turn, gives those cues increased attentional priority—that is, an addiction-related attentional bias (for reviews, see Anderson 2016; Field et al. 2016). In other words, addiction-related conditioned stimuli acquire the ability to hijack one’s attention (Field and Duka 2002; Hogarth and Duka 2006; Robinson and Berridge 2008). However, it is worth noting that contemporary perspectives on attentional bias in other clinical fields suggest that attentional bias might result from impairments in the central executive (e.g., Heeren et al. 2017; Heeren et al. 2013), stressing the fuzzy boundaries of the mechanisms underlying attentional bias.

At the empirical level, a wealth of research indicates that individuals with addictive behaviors exhibit preferential attentional allocation toward addiction-related stimuli (e.g., alcohol-related cues) rather than toward other stimuli. Notably, attentional biases have

repeatedly been shown by using various paradigms across different addictive disorders, including severe alcohol use disorders (e.g., Townshend and Duka 2001), tobacco smoking (e.g., Chanon et al. 2010), cannabis use (e.g., Vujanovic et al. 2016), cocaine use (e.g., Mayer et al. 2016), heroin addiction (e.g., Waters et al. 2012), problem gambling (e.g., Hønsi et al. 2013), and problem video gaming (Jeromin et al. 2016). Of note, there is also strong evidence that addiction-related attentional biases might be more than just a by-product of addiction. For instance, across the different addictive disorders, the magnitude of attentional biases has been shown to be strongly predictive of later relapse (Carpenter et al. 2006; Cox et al. 2002; Marissen et al. 2006; Powell et al. 2010; Waters et al. 2003). Moreover, research has shown that reducing attentional biases can, in turn, foster a cascade of downstream benefits, such as the reduction of substance use and the improvement of treatment among individuals with alcohol and nicotine dependence (e.g., Fadardi and Cox 2009; Field et al. 2009; Schoenmakers et al. 2010; but see Christiansen et al. 2015). Together, these observations thus stress the relevance of considering attentional bias as a core transdiagnostic feature of addiction.

Measuring Attentional Bias The Visual Dot-Probe Task appears to be the most commonly used tool to assess attentional biases in contemporary psychopathology research (e.g., Field et al. 2004; Harvey et al. 2004). Notably, most of the eye-tracking studies in addiction research have been performed during a dot-probe task (Field et al. 2006; Marks et al. 2014). In this task, participants view two stimuli (e.g., an alcohol-related word/photograph and a neutral word/photograph) simultaneously presented in two distinct locations (left/right or up/down) on a computer screen for a brief duration (e.g., 500 ms). Immediately thereafter, a probe appears in the location previously occupied by one of the stimuli. In different versions, participants have to indicate the location (right/left or up/down) or identity (e.g., “E” or “F”) of the probe as quickly as possible. Attentional bias is demonstrated when participants

respond faster to the probe when it replaces an addiction-related stimulus than when it replaces a neutral stimulus, indicating that their attention has been directed to the location occupied by the former stimulus.

Associative Processes: Definition and Role in Addiction Associative processes can be distinguished in accordance with the proxies that they serve, that is, on the one hand, the strength of the associative structure *per se*, and on the other hand, the automatic tendencies (Gawronski and Bodenhausen 2006; Hofmann et al. 2009; Woud et al. 2016). At a basic level, implicit associations are thought to result from (or be progressively strengthened by) temporal or spatial contiguity with external stimuli, affective reactions, and related behaviors (Sloman 1996). For instance, during the development of an addictive disorder, the related behavior becomes progressively controlled by addiction-associated information that has acquired, via classic and instrumental learning mechanisms, the property of being able to drive addiction-related actions. Moreover, implicit associations and related automatic tendencies are assumed to be independent of whether a person consciously endorses or rejects the implication of an associative link (Aarts et al. 2001; but see O'Doherty et al. 2017). In other words, although someone may be aware of the increasing negative consequences of the execution of an addictive behavior, external stimuli can automatically elicit implicit associations and automatic tendencies (e.g., Wiers et al. 2009; but see Pauli and O'Reilly 2008; Shanks 2009).

In view of the aforementioned literature, researchers have extensively studied the implicit associations and automatic approach-avoidance tendencies in addictive disorders. In particular, several lines of research have led to conceptualizing these associative processes as transdiagnostic features of addiction. First, in nonclinical populations, both implicit associations and automatic tendencies can predict later use of alcohol (Wiers et al. 2002), cigarettes (McCarthy and Thompsen 2006), and cannabis (Ames et al. 2007). Moreover, numerous studies have reported strong implicit associations and automatic tendencies among

individuals with addictive behaviors. Positive implicit associations have notably been found among alcoholic patients (Barkby et al. 2012), cigarette smokers (Payne et al. 2007; Woud et al. 2016), cannabis users (Cousijn et al. 2011), cocaine users (Wiers, Houben et al. 2007), heroin users (Wang et al. 2015), problem gamblers (Brevers, Cleeremans, Hermant et al. 2013), and problem video gamers (Yen et al. 2011). Similarly, approach and avoidance automatic tendencies have been found in alcohol dependence (Barkby et al. 2012; Field et al. 2008), nicotine dependence (Mogg et al. 2005), cannabis use (Beraha et al. 2013), and problem gambling (Boffo et al. 2018).

Of clinical importance, implicit associations and approach-avoidance tendencies are strongly predictive of future addiction-related behaviors, as well as treatment responses among alcohol-dependent individuals (Field et al. 2017; Martin Braunstein et al. 2016; Spruyt et al. 2013; Wiers et al. 2011), smokers (Chassin et al. 2010), and heroin users (Marhe et al. 2013). Moreover, modifying either implicit associations (Tello et al. 2018) or automatic tendencies (Sharbanee et al. 2014) can reduce drinking behaviors in community samples. Preliminary research has revealed similar findings among alcohol-dependent individuals (Manning et al. 2016) and cigarette smokers (Wittekind et al. 2015), stressing the pivotal role of associative processes in the maintenance of addictive disorders.

Measuring Associative Processes Traditionally, associative processes are typically assessed via implicit measurements (De Houwer 2006). Basically, these measurements can be separated in accordance with the proxies they serve, that is, on the one hand, impulse strength (i.e., the strength of the associative structure *per se*), and on the other hand, automatic tendencies reflecting the impulse (Hofmann et al. 2009). Two prominent measures of association strength are the Implicit Association Test (Greenwald et al. 1998) and the Affect Misattribution Paradigm (Payne et al. 2005). Regarding automatic tendencies toward the temptation of interest, approach-avoidance paradigms are the most popular measures (e.g.,

Wiers et al. 2011). Some of these measures involve actual approach-avoidance responses (e.g., Rinck and Becker 2007), whereas others use symbolic responses (e.g., Mogg et al. 2005).

In recent years, the Approach Avoidance Task (Rinck and Becker 2007) has become particularly popular in addiction research. This task involves the execution of approach-avoidance motor responses, requiring participants to respond to stimuli that appear on a computer screen by either pushing (avoidance movement) or pulling (approach movement) a joystick. Given this operationalization, the time participants need to fully execute these valenced movements provides an index of their automatic tendencies for a particular stimulus (e.g., alcohol versus soft drinks). Of note, most of the approach-avoidance paradigms include a task-irrelevant feature so that the push and pull movements are not determined by the content, but by the format or the tilt of the pictures (e.g., push landscape versus pull portrait pictures). As these instructions are stimulus independent, the required response is thus unrelated to the picture's content. As an illustration, it was found that, via this task, hazardedly drinking students were faster to pull than to push the joystick in response to alcohol pictures (Wiers et al. 2009). Likewise, automatic approach tendencies were positively associated with past-month expenditures and gambling frequency among problem gamblers (Boffo et al. 2018).

Despite their popularity, implicit measurements have been strongly criticized recently, especially given their relatively low reliability (e.g., Reinecke et al. 2010). Moreover, uncertainty still abounds regarding whether implicit tasks really measure automatic associations, which are assumed to cause the outcome of the tasks (De Houwer et al. 2009).

Inhibition: Definition and Role in Addictive Disorders Inhibition globally relates to the ability to control one's behaviors, thoughts, or emotions to produce contextually or socially

1 adapted actions. Inhibitory mechanisms are currently conceptualized as a constellation of
2 processes. One of the most influential models (Friedman and Miyake 2004) separates three
3 inhibition-related functions: (1) prepotent response inhibition, the ability to voluntarily stop or
4 defer a dominant or habitual response; (2) resistance to distractor interference, the ability to
5 specifically process a pertinent stimulus while reducing or suppressing the processing of
6 interfering irrelevant stimuli; and (3) resistance to proactive interference, the ability to reduce
7 or avoid the intrusion of previously relevant but now obsolete information in memory that
8 interferes with the ongoing task.

9 The loss of control over one's behavior is the hallmark of all addictive disorders, and
10 inhibition is thus at the heart of addictions, which can be defined as the reduction in the ability
11 to inhibit an approach behavior toward a substance or stimulus. In view of this pivotal role,
12 inhibition has been the most deeply explored process in addictive disorders, with recent
13 reviews (Smith et al. 2014; Spechler et al. 2016) underlining that reduced inhibition is a
14 transdiagnostic characteristic of addicted individuals. Hundreds of results from studies that
15 used various tasks and methodologies have extensively shown that the excessive use of
16 alcohol (Stavro et al. 2013), nicotine (Luijten et al. 2013), cannabis (Tapert et al. 2007),
17 cocaine (Morie et al. 2014), and MDMA (Roberts et al. 2013), as well as problem gambling
18 (Moccia et al. 2017) and problem video gaming (Littel et al. 2012), is related to a reduction in
19 the inhibition brain network and reduced performance in inhibition tasks. When separately
20 investigating the three subcomponents mentioned above, studies have shown the following:

21 1. Prepotent response inhibition is impaired with a medium-large effect size in
22 methamphetamine (Tabibnia et al. 2011) and gambling (Grant et al. 2011) disorders, as well
23 as with a small-medium effect size in cocaine abuse (Colzato et al. 2007; Fillmore and Rush
24 2002) and severe alcohol use disorders (Goudriaan et al. 2006). Conversely, heterogeneous

results have been found for nicotine and cannabis consumption (Billieux, Gay, Rochat, Khazaal et al. 2010; de Ruiter et al. 2012; Moreno et al. 2012; Ramaekers et al. 2009).

2. Resistance to distractor interference is reduced in alcohol use disorders, with a bidirectional causal link: reduced performance has been reported at the early stages of excessive alcohol use (Lannoy et al. 2017) and might constitute a vulnerability factor (van Hemel-Ruiter et al. 2015), but long-term excessive alcohol use also has a direct negative impact on this ability (Schellekens et al. 2010). Nicotine dependence (Carim-Todd et al. 2016), heroin dependence (Chen et al. 2013), and cannabis use (Maij et al. 2017) do not seem to strongly reduce resistance to distractor interference, whereas cocaine use does (Sokhadze et al. 2008).

3. Resistance to proactive interference has been little studied and related results were inconsistent, as only two studies conducted in recently detoxified alcohol-dependent individuals are available, the first showing a deficit in resistance to proactive interference (Noël et al. 2009) and the second revealing preserved performance (Noël, Van der Linden et al. 2013). These results suggest that differential patterns of deficits might be present across inhibition subcomponents in addictive disorders. However, as most studies have focused on prepotent response inhibition, the two other subcomponents are still largely underexplored in addictive disorders. Future studies should thus go beyond the measurement of general inhibition skills to promote a separate and specific exploration of each inhibition subcomponent, allowing the establishment of a far more complete profile of inhibitory performance at experimental and clinical levels.

Measuring Inhibition abilities constitute subcomponents that might be differentially altered in various addictive states and should thus be measured separately. Prepotent response inhibition can be measured by using the Stop-Signal Task (Verbruggen and Logan 2008), in which participants have to perform a simple binary decision task on targets but refrain from

responding when a sound (the stop signal) is presented right after the target (usually in 20-25% of the trials). Stopping the response requires a fast control mechanism to prevent the execution of the already planned motor response. The time needed to stop a response (stop-signal reaction time, or SSRT) is covert, as no behavioral response is given for efficient inhibition. It is thus estimated by varying the time between stimulus presentation and stop-signal occurrence across trials and by computing the probability of the participant stopping at different stop-signal delays. This individual SSRT is the main dependent inhibition measure.

Resistance to distractor interference can be measured by using the Flanker Task (Eriksen and Eriksen 1974), in which participants have to make a binary decision on a central letter while ignoring surrounding and irrelevant letters (i.e., the flankers). This task compares experimental-incongruent conditions (where flankers promote the opposite response to that requested by the central letter) and control-congruent conditions (where flankers promote the identical response to that requested by the central letter). Reduced performance (reduced accuracy or increased reaction times) in the incongruent condition indexes the interference effect from the distractor flankers.

In view of the limited evidence currently available concerning the presence of deficits in the resistance to proactive interference in addictive disorders, no task related to this process has been included in the proposed assessment battery.

Flexibility: Definition and Role in Addictive Disorders Cognitive flexibility relates to the skills needed to rapidly adapt one's thoughts or behaviors following environmental changes. This ability partly relies on other high-level functions (Miyake and Friedman 2012) and has, moreover, been dissociated in several subcomponents according to task requirements, notably leading to the distinction between set shifting, intradimensional flexibility, and extradimensional flexibility (McAlonan and Brown 2003). For the sake of clarity, this

executive subcomponent will here be conceptualized as centrally related to shifting abilities, namely, the capacity to switch back and forth between different tasks or mental sets (Monsell 1996).

Addictive disorders are related to flexibility impairments: addicted individuals are constantly using the same behavioral pattern (i.e., addiction-related behavior) and are unable to flexibly adapt their actions according to changing contexts or predicted outcomes. The importance of impairments in flexibility in addictive behaviors has been widely recognized, with reduced task or mental set shifting abilities documented in alcohol-related disorders (Czapla et al. 2016), cocaine users (Van der Plas et al. 2009; Woicik et al. 2009), and gambling disorder (Leppink et al. 2016). The use of tasks more specifically measuring cognitive flexibility, such as reversal learning paradigms (Cools et al. 2002), showed that patients with severe alcohol use disorders need more time to extinguish previously learned contingencies, indexing reduced flexibility (Vanes et al. 2014). In line with this, cocaine (Verdejo-García et al. 2015) and cannabis (Spronk et al. 2016) users also present increased flexibility errors, which are also found in nicotine-dependent patients currently experiencing craving (Lesage et al. 2017). The use of such specific tasks led to contradictory results in gambling disorders, three studies showing impaired cognitive flexibility (Boog et al. 2014; Choi et al. 2014; de Ruiter et al. 2009) and two studies not reporting such differences with control groups (Torres et al. 2013; Verdejo-García et al. 2015). Finally, Choi et al. (2014) found that problem video gamers present performances that are similar to those of healthy controls in tasks assessing cognitive flexibility. Notably, flexibility deficits play a direct role in the maintenance of disease by constituting a crucial predictor of treatment outcomes in cocaine users (Turner et al. 2009) and alcohol-dependent individuals (Desfosses et al. 2014).

Measuring Flexibility In the Reversal Learning Task (Cools et al. 2002), each trial presents two abstract stimuli and the participant has to make a choice. Feedback is given (“good” or

“bad”), and the participant progressively learns which option is good or bad. The predictive value of the stimulus for the outcome is probabilistic (i.e., the feedback is correct in 80% of trials). After efficient learning, a rule is modified (i.e., a reversal) and the participant has to learn the new rule. Dependent measures are the number of trials needed to learn this new rule, indexing cognitive flexibility. While other efficient tasks might be used to measure flexibility in addictive disorders (e.g., Intra/Extra Dimensional Set Shifting Task, Roberts et al., 1988), this task has the central advantage of efficiently measuring cognitive flexibility in an ecological design while allowing dissociation of a shifting component (perseverative errors, responding to the previously reinforced stimuli despite negative feedback) from a negative-feedback component (probabilistic errors, responding correctly but receiving negative feedback).

Updating: Definition and Role in Addictive Disorders Updating refers to the encoding and monitoring of stimulations to modify short-term memory traces according to task requirements (Miyake et al. 2000). It is thus related to working memory (Baddeley 1986; Diamond 2013) but is considered an executive function that goes beyond memory abilities (Friedman and Miyake 2017), as it implies controlled use of encoded stimulations (i.e., manipulating the memorized content).

The effects of drugs on updating efficiency have long been suggested (Fattore and Diana 2016; Stavro et al. 2013): decreased ability to update memory traces reduces everyday life efficiency and could play a role in the persistence of addiction-related habits. A large range of updating paradigms have been used in addictive disorders, indicating strongly reduced performances. In severe alcohol use disorders, updating constitutes a key impairment in the weeks following detoxification, at behavioral (reduced performance) and cerebral (reduced dorsolateral prefrontal cortex activations) levels (Pfefferbaum et al. 2001), whereas heavy consumers without alcohol dependence might present compensatory brain strategies

resulting in preserved performance (Wesley et al. 2017). The clinical importance of these brain activations related to updating has been confirmed by results showing that increased prefrontal activations during updating tasks at treatment initiation predict sustained abstinence (Charlet et al. 2014). A long-term impact of drug consumption on updating has also been found in MDMA (Verbaten 2010), cocaine (Verdejo-García and Pérez-García 2007), and heroin (Papageorgiou et al. 2003) dependence, as well as in cannabis consumption, where updating impairments constitute a hallmark of acute (Keles et al. 2017) and chronic (Cohen et al. 2017) use. Regarding nicotine consumption, the results are less clear, as acute nicotine consumption increases updating abilities and long-term dependence is related to impaired updating (Valentine and Sofuoglu 2018), this deficit predicting early relapse (Patterson et al. 2010). Finally, preliminary work shows that updating might be affected in gambling disorder (Ledgerwood et al. 2012), but few reliable results are currently available with regard to behavioral addictions.

Measuring Updating Classic updating tasks all require active processing of freshly encoded information (Diamond 2013). In particular, the N-Back Task (Kirchner 1958) is considered a reliable tool to measure verbal updating abilities with strong control of working memory load (Friedman and Miyake 2017). This task is usually based on visual stimuli: letters (or numbers) are presented one by one on the screen, and for each stimulus, the participant has to determine whether it is identical or different from the one presented “n” trials before, “n” varying across conditions and studies between 0 (i.e., detecting the appearance of a prespecified stimulus stored in memory without an updating requirement) and 3 (i.e., deciding whether the presented stimulus is identical to the stimulus presented three trials earlier). Comparing the performances (percentage of correct detections) between control conditions (0-back, 1-back) and experimental conditions (2-back or 3-back) leads to a specific updating

index, namely, the ability to monitor, maintain, and modify the information contained in working memory.

Mental Time Travel: Definition and Role in Addictive Disorders Past- and future-oriented mental time travel constitutes a central component of human decision making. Indeed, making appropriate decisions in a specific situation requires both retrieving past similar experiences stored in episodic memory, including their emotional content (see the somatic markers hypothesis; Damasio 1996), and/or temporarily disengaging from the immediate environment to mentally explore possible futures (Boyer 2008; Noël, Jaafari et al. 2017). For instance, in the case of poor future-oriented mental time travel abilities, short-term rewards (e.g., consuming a drug to relieve negative mood) are given priority over the avoidance of long-term negative consequences of a choice or the accomplishment of future goals associated with a higher reward.

Not only is episodic memory involved in the retrieval and re-experience of past memories, but it also plays a significant role in imagining future events (e.g., Addis et al. 2007). Therefore, both past and future event representations can be episodic, comprising abundant contextual details about events that are specific in time and location (Addis et al., 2007). Mental time travel also involves executive processes and visual-spatial processing (D’Argembeau et al. 2010). Notably, mental time travel is supplemented with “cognitive feelings” (Conway 2009), such as autonoetic consciousness, which is the subjective feeling of re-experiencing past events and the feeling of being brought forward in time to pre-experience a future event (Tulving 2002). Autonoetic consciousness for future events may play an important motivational role in achieving imagined future events (e.g., abstinence), thereby constituting a crucial determinant of change in patients with addictive disorders (Noël, Jaafari et al. 2017).

Episodic memory as assessed by classic tasks in which participants have to remember a list of words or pictures has been found to be frequently impaired in patients with various addictive behaviors (see Noël, Jaafari et al. 2017). However, examining mental time travel and associated cognitive feelings requires the use of more specific procedures that assess various dimensions of autobiographical memories and future prospects. In specific procedures that assess past oriented mental time travel, recently detoxified patients with severe alcohol use disorders were impaired in recalling specific autobiographical memories in comparison to matched healthy controls (Cuervo-Lombard et al. 2016). Furthermore, this overgeneralization of autobiographical memories persists for 2 to 6 months of sobriety (Nandrino et al. 2016; Nandrino and Gandolphe 2017). Alcoholics also showed poorer episodic autobiographical performances across several life periods (Nandrino et al. 2016). These reduced performances in alcoholism could be explained by the deleterious effect of alcohol on memory consolidation, by a strategy that decreases distress associated with the retrieval of a memory, or by executive deficits that hamper strategic encoding and retrieval (Noël, Jaafari et al. 2017).

Regarding episodic future thinking, Mercuri et al. (2015, 2016) showed that opioid users tended to generate fewer episodic details when imagining novel future scenarios in response to either a positive, neutral, or negative cue word than did healthy controls, but did not differ in episodic memory, nor in the scene construction or narrative ability conditions. Preserved episodic memory but impaired episodic foresight ability suggests that intact episodic memory functioning is not sufficient for episodic foresight (Mercuri et al. 2015). In another study, Noël, Saeremans et al. (2017) showed that problem gamblers' future thoughts contained fewer episodic details (i.e., sensory details and contextual information) and were associated with reduced autonoetic consciousness compared with those of healthy controls. In addition, their mental representations of future events were less differentiated as a function of

temporal distance than were those of healthy controls. Indeed, in contrast to healthy controls, problem gamblers did not report more sensory details for closer than distant future events and their subjective sense of the temporal proximity of imagined events was less variable across temporal distances (Noël, Saeremans, et al. 2017).

These results open up a relevant prospect for understanding addiction-related treatment outcomes and symptoms (e.g., relapse, treatment adherence and outcome, readiness to change, craving, or loss of control). In particular, relapse prevention strategies (e.g., developing skills to assertively decline future drug proposals) precisely require abilities to project oneself into the future. Taking into account impairments in episodic future thinking may thus clarify why achieving and maintaining abstinence is frequently so challenging for some patients (Mercuri et al. 2015, 2016).

Measuring Mental Time Travel The autobiographical interview (Levine et al. 2002; adapted version by Addis et al. 2008) is a semistructured interview that provides an index of both episodic memory and episodic foresight. More specifically, participants are instructed to describe a previously experienced past event or a novel future event in response to a cue word (neutral, positive, or negative). The events have to refer to a specific time and place, last less than 1 day, and be described from the participant's subjective perspective. The number of episodic details specific to the central events generated for past and future events is the primary measure of episodic memory and episodic future thinking, respectively. Of note, this assessment can be complemented by adding specific questions related to the phenomenological characteristics of the imagined events and to the subjective feelings (autonoetic processes) associated with the events (D'Argembeau and Van der Linden 2006).

Interoception: Definition and Role in Addictive Disorders Interoception refers to a range of processes by which body-relevant signals are transmitted to the brain, allowing the organism to regulate internal states and maintain homeostasis (Verdejo-García et al. 2012).

Interoception is underlain by multiple processes, including (1) interoceptive accuracy, which typically refers to performance on objective behavioral tests (e.g., heartbeat detection); (2) interoceptive sensibility, which refers to subjective interoception as assessed by self-reports; and (3) interoceptive awareness, which refers to a metacognitive dimension of interoceptive accuracy, as assessed by confidence-accuracy correspondence in a performance task. It thus reflects whether an individual knows if he or she is making good or bad interoceptive decisions with respect to interoceptive behavioral accuracy (Garfinkel et al. 2015).

Neuroimaging studies have consistently stressed that the anterior insula is the critical hub of interoceptive processes (Critchley et al. 2004). According to several authors, disturbances in the prediction or simulation of the body by the insular cortex constitute a key aspect of addiction (Naqvi and Bechara 2010; Naqvi et al. 2007; Paulus et al. 2009; Verdejo-García et al. 2012). Indeed, the insula associated with various regions of the prefrontal cortex (ventromedial and orbital) conveys both the conscious pleasure resulting from the bodily effects of drug taking (e.g., the airway effect after a cigarette puff) and the conscious pain associated with drug withdrawal. In addition, the insula may be involved in the recall and simulation of these bodily effects, which promotes conscious cue-induced urges, desires, or anticipation (e.g., when seeing someone drinking a Belgium beer in a bar), as well as in a cost-benefit analysis of the interoceptive effects of drug consumption. For instance, in the context of failed abstinence, the positive effect of drug taking encoded in the insula (e.g., pleasure) outweighs the negative effects (e.g., aversion). Urge or craving due to heightened insular activity may increase compulsive drug seeking while undermining cognitive resources necessary for adapted decision making (Noël, Brevers et al. 2013a, 2013b).

Corroborating this view, many studies have stressed the positive correlations between insula activity and self-reported urge for substances such as cigarettes, cocaine, alcohol, or heroin (see Verdejo-García et al. 2012), as well as higher levels of nicotine dependence

(Goudriaan et al. 2010). Conversely, damage to the insular cortex tends to dramatically reduce the intensity of craving episodes and provoke immediate smoking cessation (without relapse) in cigarette consumers (Naqvi et al. 2007). However, a decrease in interoceptive function associated with addiction should also be considered (Verdejo-García et al. 2012). Indeed, in such cases, the insular cortex may not appropriately signal the potential aversive outcome of a choice in risky situations (Paulus and Stewart 2014), allowing individuals who are influenced by non-emotional sources such as habits to be guided in their decision making.

By using laboratory tasks that assess interoceptive accuracy, some studies conducted on healthy adults from the community showed that cardiac interoceptive ability critically influences decision making and loss aversion on laboratory decision-making tasks (Dunn et al. 2010; Sokol-Hessner et al. 2015; Werner et al. 2009, 2013). In clinical samples, sober substance use disorder patients (alcohol, heroin, or synthetic cannabinoids) showed significantly lower interoceptive cardiac accuracy than did healthy controls (Ates Cöl et al. 2016; Sönmez et al. 2017). In addition, this poor interoceptive accuracy was significantly associated with greater alcohol craving and greater difficulties in identifying feelings (a dimension of alexithymia), but not with disorder severity. Finally, Schmidt et al. (2013) showed in a sample of in- and outpatients treated for alcohol disorder or multiple substance use disorders (including alcohol) that interoceptive accuracy and tension reduction motives (e.g., negative reinforcement-based expectancies) interacted to predict drinking-related obsessions and compulsions.

Measuring Interoception Most procedures have been developed to assess interoception accuracy in the cardiac domain for pragmatic reasons, inasmuch as heartbeats are distinct and frequent internal events that can be easily measured. The most popular procedures are the Heartbeat Detection Task and the Mental Heart Beat Tracking Method. In a Heartbeat Detection Task (Barrett et al. 2004), during each trial, participants hear a series of 10 tones,

each of which is generated by the R-spike in their electrocardiogram. The participants are then asked to report on whether the series of tones are coincident or not with their own heartbeats. On coincident trials, the tones are in general presented 200 ms after each R-spike, whereas they are presented 500 ms after each R-spike on noncoincident trials. In a mental tracking method, participants are asked to count the number of heartbeats experienced in short time intervals (e.g., 25-55 s). The experimenter then compares participants' estimation of their own heartbeats with the actual number of heartbeats recorded on an electrocardiogram trace. Of note, recent data showed that the mental tracking method is strongly affected by expectations and beliefs about one's own heartbeat and should thus be avoided (Ring et al. 2015).

Self-report questionnaire measures of interoception, such as the awareness section of the Body Perception Questionnaire (Porges 1993), are useful in measuring individual differences in perceived sensitivity across a wide range of internal bodily changes, but they do not inform about subjective interoceptive accuracy (Garfinkel et al. 2015). This subscale contains 45 items referring to various bodily sensations (e.g., stomach and gut pains, respiration, heartbeat). For each item, participants indicated their awareness of each sensation on a five-point scale ranging from 1 ("never") to 5 ("always"). Finally, interoceptive awareness can be quantified by measuring the degree to which the accuracy of the performances on a Heartbeat Detection Task is predicted by subjective confidence in the task judgement (Garfinkel et al. 2015). For this purpose, following each trial on a Heartbeat Detection Task, participants are asked to immediately rate their confidence in the perceived accuracy of their response on a continuous visual analogue scale ranging from "Total guess/No heartbeat awareness" to "Complete confidence/Full perception of heartbeat."

Self-awareness: Definition and Role in Addictive Disorders It has been proposed that dysfunction of the interoceptive system not only influences the processing of drug-related

cues, but that it likely disturbs self-awareness, which may result in underestimation of the severity of a disorder, failure to recognize an illness (or lack of insight), denial, or dissociation between intention and action (Goldstein et al. 2009; Goldstein and Volkow 2011; Verdejo-García et al. 2012) when drug-induced damage to the insula interrupts the interoceptive input signaling the current state of the body. Lack of insight is an important issue in addiction treatment in that some individuals who require treatment probably do not recognize the need for therapeutic support. This reduced perceived need for seeking treatment and/or underestimation of addiction severity may result in poor decision making in the context of continuous substance misuse or addictive behaviors despite repetitive negative outcomes (Brevers and Noël 2013; Goldstein et al. 2009; Goldstein and Volkow 2011).

In the field of addictive disorders, most studies that aimed to examine self-awareness used procedures measuring retrospective metacognitive ability such as rating confidence or wagering on one's own decision, which allows dissociation of metacognition from objective task performance on a trial-by-trial basis. Nevertheless, such procedures examine only one specific aspect of self-awareness, that is, to which point people can monitor their own decision making and discriminate between accurate and inaccurate judgments (Moeller et al. 2016). By using such procedures, dissociations between self-perception of performances and actual behavior have been found in cocaine users, with implications for increased drug-seeking behavior and decreased social-emotional functioning (Moeller et al. 2010, 2012, 2014, 2016); in patients with nicotine dependence (Chiu et al. 2008); in former opium and heroin users who are stable on methadone maintenance treatment (Sadeghi et al. 2017); and in disordered gambling (Brevers, Cleeremans, Bechara et al. 2013; Brevers et al. 2014). In gambling disorder, this overconfidence in one's bet or decision is also influenced by perceived control, which results in reduced betting performance (Goodie 2005).

Measuring Self-awareness in Addiction In the Visual-Perception Task developed by Moeller et al. (2016), participants are instructed to make two-choice discrimination judgments about what they have just perceived before providing a confidence rating on their decision. More specifically, on each trial of this task, two circles are presented on the computer for 1 s. A variable number of dots are then displayed inside both circles for 0.7 s. The aim of the task is to guess whether the left or the right circle contained more dots. The difference in the number of dots between the two circles is titrated so that each participant's performance is maintained at a constant level. After an incorrect response, the difference in the number of dots between the two circles is increased by one unit, whereas after two consecutive correct responses, the difference in the number of dots is decreased by one dot. Using a staircase procedure permits one to equate the difficulty of the task between individuals. In total, each participant completes 200 perception trials. To estimate metacognitive efficiency, the investigators compute meta- d' (Maniscalco and Lau 2012). In a signal detection theory, meta- d' is considered a measure of type 2 sensitivity (i.e., the degree to which individuals can discriminate their own correct judgments from their own incorrect judgments).

Social Cognition: Definition and Role in Addictive Disorders Social cognition is a broad concept encompassing the psychological resources needed to correctly identify interpersonal signals and the skills needed to effectively respond to these signals. This ability is constantly needed in everyday life to efficiently detect, understand, regulate, and react to stimulations emerging from the interpersonal environment (Green et al. 2015). To clarify this field, Green et al. (2008) proposed a classification of social cognition subcomponents, identifying five subcategories: (1) *theory of mind*, the ability to use interpersonal signals to infer others' mental states and to anticipate their behaviors; (2) *social perception*, the aptitude to interpret verbal and nonverbal stimuli to infer individuals' roles and current relationships in equivocal interpersonal contexts; (3) *social knowledge*, the awareness and understanding of the social

rules and conventions; (4) *attributional bias*, the propensity to consider that interpersonal events are mostly due to individual responsibility, or conversely to others' responsibility or contextual variables; and (5) *emotional processing*, the efficient perception, interpretation, and reaction to the emotional states expressed by others through face, voice, or posture. This typology offers a sound classification to clarify the studies conducted in addictive disorders.

However, as mentioned earlier, the exploration of these social cognition subcomponents is still poor in addictive disorders. Most studies exploring social cognition abilities in addicted populations have focused on severe alcohol use disorders and have led to the proposal of a global social cognition deficit (Bora and Zorlu 2017; Onuoha et al. 2016). Although attribution biases have not yet been explored, theory of mind (Bosco et al 2014), empathy (Martinotti, et al. 2009), social perception (Maurage et al. 2016), social knowledge (Thoma et al. 2013), and emotional processing (Donadon and Osório 2014) are impaired in this population. Conversely, the exploration of social cognition in other substance-related addictive states has been very limited. The only available data suggest that, although nicotine dependence does not impact social cognition (Drusch et al. 2013), reduced theory of mind abilities have been documented in cocaine abusers, this deficit being correlated with craving intensity (Sanvicente-Vieira et al. 2017). Moreover, despite strong interstudy heterogeneity, substance-related addictions are reliably associated with reduced emotion perception abilities (Castellano et al. 2015): individuals who are dependent on cannabis (Bayrakçı et al. 2015), cocaine (Ersche et al. 2015), methamphetamine (Henry et al. 2009), heroin (Contero et al. 2012), or polysubstances (Fernández-Serrano et al. 2010) present reduced emotional facial recognition performances. Finally, social cognition has been explored in only two studies in gambling disorder, which suggested reduced empathy (Tomei et al. 2017) and emotional decoding (Kornreich et al. 2016) abilities. In conclusion, and while only preliminary evidence is available, social cognition deficits may constitute a core characteristic of addictive

disorders. Their suggested causal role in the development of addictive disorders (Ernst et al. 2010) and in relapse after detoxification (Zywiak et al. 2003) should urge researchers to develop a theoretically based exploration of these abilities across addicted populations.

Measuring Social Cognition Four tasks can be proposed, each specifically related to a subcomponent of the typology presented above (Green et al. 2015). Because attributional biases have not yet been measured in addictive disorders, we decided not to include a task measuring this process in our battery.

1. *Theory of mind* can be evaluated by using a nonverbal and video-based version of the False Belief Task (Apperly et al. 2004), which measures the ability to infer others' mental states to predict their behaviors. In this task, the participant is asked to make judgments regarding which of the two identical boxes presented in the video contains a green cube by using cues given by actors. In a first subtask ("track-task"), the participant has to determine which box contains the green cube by tracking the actor's perspective and taking into account that this actor has, in some trials, false beliefs about this location (leading him to give erroneous cues). In the second subtask ("inhibit-task"), the participant has to determine which box will be chosen by the actor, which requires the participant, for some trials, to inhibit his or her own correct knowledge about the object's location (self-perspective inhibition) in order to infer the actor's behavior resulting from his false belief. This task is a sound and specific measure of theory of mind, which assesses the ability to understand others' mental states through inhibition of self-perspective.

2. *Social perception* can be evaluated by using the Movie for Assessment of Social Cognition (Dziobek et al. 2006), which ecologically evaluates the perception of social signals by asking the participant to identify the emotions, thoughts, and intentions expressed by characters in 45 short videos depicting real-life social interactions. During each video, social interactions are

depicted through characters that express positive or negative emotions, feelings, intentions, and thoughts with different intensities. After each video, a multiple-choice question is asked to evaluate the participant's ability to perceive the affective or cognitive social signals expressed. Each answer corresponds to a response type, related to correct, excessive, reduced, or absent social perception, this task thus measuring the ability to efficiently perceive and interpret social signals.

3. *Social knowledge*, which is centrally related to semantic and rule-based skills, can be evaluated by using the Faux-Pas Test (Stone et al. 1998), which requires the participant to identify the breaking of social rules. Twenty written stories are presented, half of them involving an interpersonal mistake. For each story, the participant is asked whether a social faux-pas was present and, if so, which character committed the faux-pas. This task measures the level of social knowledge by measuring the participant's ability to detect interpersonal mistakes in social contexts.

4. *Emotional processing* can be evaluated with the Facial Emotion Recognition Task (Gaudelus et al. 2015), a validated emotion decoding task in which the participant has to identify the emotional state (i.e., anger, contempt, disgust, fear, happiness, sadness) presented by human faces. Faces are morphed (from neutral to full-blown emotion) to evaluate nine levels of intensity (i.e., from 20 to 100%), leading to a total of 54 stimuli (6 emotions $\times$ 9 intensities). The accuracy score and detection threshold (intensity at which the emotional content is reliably detected) are computed for each emotion.

Discussion

This paper aimed to describe the core features of decision making in addictive behaviors by proposing a range of processes, with a related measurement task, initiating a

comprehensive process-based decision-making assessment battery. Table 1 summarizes the selected processes and the typically related measurement tasks.

--INSERT TABLE 1 HERE--

The new and innovative processual and transdiagnostic theoretical background proposed here could account for various outcomes related to poor decision making across addictive disorders, but could also, from a transdiagnostic perspective, be applied to other psychopathological disorders. By identifying core processes involved in decision making, the present work thus contributes to enriching the understanding of what constitutes (impaired) decision making, ultimately sophisticating existing system-based (i.e. dual or triadic) models of decision making that still theoretically anchor a large body of experimental and applied research.

According to the conceptual framework proposed, the various processes retained here may contribute to a better understanding of the cognitive steps involved in decision making as conceptualized by Verdejo-García et al. (2018), namely the formation of preferences, choice implementation, and feedback processing and learning. First, formation of preferences is likely to be influenced by various processes, including associative processes (susceptible to promoting automatic approach tendencies), attentional bias (toward salient conditioned addiction-related cues), interoceptive factors (susceptible to promoting craving through insular activity), reward sensitivity (preference for rewarding, risky, or stimulating actions), and poor self-awareness (susceptible to preventing one from discriminating between accurate and inaccurate judgments). Second, we propose that choice implementation primarily capitalizes on the efficacy of executive processes (i.e., prepotent response inhibition or flexibility) to restrain and override temptation and to inhibit irrelevant information or actions. Other processes such as mental time travel and self-awareness probably also play a pivotal role in mobilizing motivational resources and proactive pursuit of selected choices. Finally,

1 efficient feedback processing relies on various processes such as executive functions (e.g.,
2 inhibition, updating, and flexibility are necessary to avoid reproducing hazardous or
3 suboptimal choices), sensitivity to reward (resulting in greater attention to gains than to
4 losses), or mental time travel (necessary to take into account past experiences in order to
5 guide further choices).

6 Another relevant aspect of our approach is that it extends previous work by including
7 up-to-date social cognition processes that have been established as playing a pivotal role in
8 disadvantageous decision making. Indeed, dual- or triadic-process models have largely
9 neglected other relevant systems involved in the emergence and maintenance of substance
10 abuse and addictive disorders. We thus decided to include social cognitive processes, as they
11 also play a critical role in the onset and perpetuation of addictive disorders (Bora and Zorlu
12 2017; Onuoha et al. 2016). However, further research is needed to disentangle the relations
13 between the social cognition-related processes and the other processes included in the
14 proposed battery. In particular, not all of these processes can be considered at the same
15 epistemological level. Indeed, on the one hand, interoceptive processes play an important role
16 in emotional processing and executive functions (see Adolfi et al. 2017; Khalsa et al. 2018;
17 Poppa and Bechara 2018). On the other hand, social cognition refers to an overarching
18 category of processes involved in perceiving others and interacting with them that is under the
19 influence of more basic processes (e.g., interoceptive processes, associative processes,
20 executive functions, attentional biases, self-awareness). An in-depth analysis of the potential
21 interactions between the social cognition processes and the other processes involved in
22 decision making is thus warranted to better appraise decision making in social contexts, that
23 is, when the behavior of others affect someone's well-being, when someone has to make a
24 choice in reference to another person, or when the respect of social norms has to be balanced
25 in regard to someone's own interests (Ruff and Fehr 2014).

1 A central aspect of further elaboration of the theoretical framework proposed would be to
 2 investigate the interactions and reciprocal influences of the distinct processes involved in
 3 decision making. In particular, to best capture the dynamic interplay among the distinct
 4 features (i.e., how they may trigger each other over time), one would need to apply
 5 computational methods that characterize the within- and between-person temporal dynamics
 6 of intensive intraindividual time series data denoting these systems and their related features
 7 (Epskamp et al. 2018). A computational network approach could also help researchers to
 8 empirically test whether the distinct features cohere as a large unitary network system of
 9 interacting elements (for illustrations, see Beveridge and Shan 2016; Heeren et al. 2018).
 10 Finally, techniques from the study of sudden transitions in ecosystems (e.g., Hirota et al.
 11 2011) may also help identify when the dynamic interplay among the distinct features is on the
 12 brink of tipping into a disordered state (e.g., disorder, relapse) or returning to a mentally
 13 healthy one. The combination of these distinct analytical approaches would ultimately
 14 conspire to elucidate the temporal dynamic interactions among the distinct features of
 15 impaired decision making in addiction and allow one to move beyond the categorical dual-
 16 and triadic approaches.

17 Our paper also opens relevant perspectives at a clinical level. Indeed, deficits in
 18 decision making are known to play a crucial role in the onset, maintenance, and relapse of
 19 substance use and addictive disorders (e.g., Bechara 2005; Bickel et al. 2018; Goudriaan et al.
 20 2005). However, taking into account the limited efficacy of current treatments, low
 21 compliance rate, and high relapse rate observed in addictive disorders (McLellan et al. 2000;
 22 Miller 1996; Moos and Moos 2006), there is an urgent need for alternative assessments and
 23 individualized, tailored, and transdiagnostic intervention strategies to prevent or diminish the
 24 personal, social, and economic burden associated with addiction. In this context, the
 25 processual and transdiagnostic battery proposed here, although preliminary, holds relevant

prospects for the diagnosis and case conceptualization, prevention, and empirically grounded treatment of decision-making impairments. Indeed, developing such a battery is all the more important because, although poor decision making may result in similar outcomes (e.g., relapse, poor treatment adherence, loss of control), the underlying processes involved in it could be heterogeneous from one person to another. Appraising the heterogeneity of the processes involved in poor decision making enables clinicians to identify individuals at risk of addictive disorders and to provide more efficient prevention strategies. In addition, such an approach supports the development and validation of individualized, tailored, and transdiagnostic treatments targeting specific psychological (or in some cases physiological) processes underlying symptoms and problematic behaviors, as well as claims for abandoning standardized treatments targeting discrete syndromes, which are still largely used in the treatment of addictive disorders (Billieux et al. 2015; Dudley et al. 2011).

As an illustration, and as emphasized earlier, it is generally acknowledged that patients with substance use or gambling disorder present with prepotent response inhibition impairment (Goudriaan et al. 2004; Smith et al. 2014). In such a context, it is tenable to defend the position that the standardized treatment of addiction should comprise a specific module or intervention targeting the improvement of inhibitory control, as most addicted patients are supposed to present this specific deficit. It is worth noting, however, that most supporting evidence in the field has been gathered via case-control studies that by definition focus on group differences and neglect the heterogeneity that potentially characterizes patients. Yet, similar to concerns of other mental conditions such as schizophrenia (Larøi and Van der Linden 2013), it may be that intradiagnostic symptom heterogeneity is more the rule than the exception and that individualized remediation is required. Although few studies tackled this heterogeneity issue in clinical settings, an interesting observation can be derived from a study that measured impulsivity-related functions (including prepotent response

inhibition) in a sample of treatment-seeking patients with gambling disorder (Billieux et al. 2012). Indeed, although this study successfully reproduced classic group-control differences in inhibitory control, it also emphasized that only a subgroup of gambling disorder patients (40%) are characterized by prepotent response impairment when an individual profile analysis is applied, which calls into question the position that inhibitory control is systematically impaired and should be treated in gambling disorder. This kind of data, in line with the process-based approach defended in the current article, supports the crucial need to profile addictive disorder patients with a theoretically grounded assessment battery that allows one to consider the inherent heterogeneity of decision-making impairments as a premise of treatment conceptualization.

The preliminary battery that we proposed thus represents a first attempt to specifically assess decision-making processes, one that goes beyond the use of traditional and multidetermined decision-making laboratory tasks such as the Iowa Gambling Task, the Game of Dice Task, the Balloon Analogue Risk Task, or delay discounting tasks. However, most of the tasks presented here have been generally used for research, and their clinical adaptability/efficiency should be confirmed. In addition, although most of the described tasks efficiently discriminate patients with addictive disorders from healthy controls, the relation to measures of daily life and relevant clinical outcomes (e.g., relapse, abstinence, treatment adherence, quality of life) have not systematically been established. Furthermore, appraising decision-making impairments in addictive disorders requires adoption of a more holistic perspective. Indeed, psychiatric disorders are considered to result from complex interactions between neural systems, psychological states, environmental inputs, and sociocultural determinants (Kendler et al. 2011). In this context, poor decision making, associated with various dysfunctional neurocognitive processes that depend on identified specific neural networks (e.g., insular cortex and interoceptive awareness), has been described in the current paper. However, this

perspective on decision making should be integrated in a larger clinical view of addictive disorders, including interactions between the processes described in the inventory of laboratory tasks and emotional factors (e.g., negative mood, anxiety), personal identity (e.g., self-efficacy), or relational factors. In a processual and integrative approach, the interactions between neurobiological factors, social factors, and life events could contribute to poor decision making through their joint effects on various cognitive, affective, and motivational processes identified in the current manuscript (Kinderman and Tai 2007).

Finally, our approach presents some limitations. First, we used a narrative methodology, implying that the information provided was not subject to standardized information selection and specific assessment criteria. Indeed, given the nature of the current article and the vast number of constructs covered (i.e., the range of cognitive, affective, and motivational processes considered in relation to decision making), a systematic and/or meta-analytic approach was not feasible. The selection of constructs retained resulted from a consensus reached by the authors and is thus inherently characterized by a certain degree of subjectivity. An alternative approach could have been to rely on a technique such as the Delphi technique, which is a structured method used to achieve consensus through an iterative process of gathering opinions from a panel of representative experts (Jorm 2015; Yücel et al. 2018). Applying a Delphi approach to the constructs retained (and related assessment tasks) could constitute a relevant further step to test and strengthen the theoretical framework developed in the current paper.

Second, the direction of causality between some processes and addictive features has not always been properly established (e.g., for updating, cognitive flexibility, mental time travel, self-awareness). Consequently, there is suggestive evidence only for the role of various specific processes in the development of addiction.

1 **Conclusion**

2 Addictive disorders constitute a major public health issue and a worldwide economic burden,
3 further increased by the low efficiency of current treatment proposals. It is now acknowledged
4 that poor decision making plays a pivotal role in these disorders, and the precise identification
5 of processes underlying poor decision making in addictive disorders is thus much needed to
6 promote effective prevention approaches and treatment options. In this regard, it is essential
7 to go beyond the use of general and multidetermined decision-making laboratory tasks and to
8 identify specific impairments in decision-making processes to foster psychological
9 interventions that are tailored to the person's specific dysfunctions. This paper, by identifying
10 specific processes involved in poor decision making, constitutes a first step toward a
11 theoretically grounded and experimentally valid evaluation of decision-making impairments
12 and rehabilitation in substance use and addictive disorders.

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Table 1. A process-based decision-making assessment battery

Processes	Measures
Reward sensitivity	Learning Reinforcement Task (Frank et al. 2004)
Attentional bias	Visual Dot-Probe Task (Field et al. 2004)
	Implicit Association Test (Greenwald et al. 1998)
Associative processes	Affect Misattribution Paradigm (Payne et al. 2005)
	Approach Avoidance Task (Wiers et al. 2011)
Inhibition	
Prepotent response inhibition	Stop-Signal Task (Verbruggen and Logan 2008)
Resistance to distractor interference	Flanker Task (Eriksen and Eriksen 1974)
Flexibility	Reversal Learning Task (Cools et al. 2002)
Updating	N-Back Task (Kirchner 1958; Friedman and Miyake 2017)
	Adapted Autobiographical Interview, including assessment of
Mental time travel	autonoetic consciousness (Addis et al. 2008; D'Argembeau and Van der Linden 2006)
Interoceptive accuracy	Heartbeat Detection Task (Barrett et al. 2004)
Interoceptive sensibility	Awareness section of the Body Perception Questionnaire (Porges 1993)
Interoceptive awareness	Heartbeat Detection Tasks with confidence rating (Garfinkel et al. 2015)
Self-awareness	Visual-Perception Judgement Task with confidence rating (Moeller et al. 2016)
Social cognition	
Theory of mind	False Belief Task (Apperly et al. 2004)
Social perception	Movie for Assessment of Social Cognition (Dziobek et al. 2006)
Social knowledge	Faux-Pas Test (Stone et al. 1998)
Emotional processing	Facial Emotion Recognition Task (Gaudelus et al. 2015)

Note. The measures proposed in this battery should be considered only as exemplars of tasks or self-reports that could be used to assess each process.