

# Digital phenotyping and mobile sensing in addiction psychiatry

Shuyan Liu<sup>1\*</sup>, Stephan Heinzl<sup>2</sup>, Raymond J. Dolan<sup>3,4,5</sup>

<sup>1</sup>Department of Psychiatry and Psychotherapy, Charité – Universitätsmedizin Berlin (Campus Charité Mitte), Berlin, Germany

<sup>2</sup>Clinical Psychology and Psychotherapy, Department of Education and Psychology, Freie Universität Berlin, Berlin, Germany

<sup>3</sup>Max Planck Centre for Computational Psychiatry and Ageing Research & Wellcome Centre for Human Neuroimaging, University College London, London, England.

<sup>4</sup>Wellcome Centre for Human Neuroimaging, University College London, London, UK

<sup>5</sup>State Key Laboratory of Cognitive Neuroscience and Learning, IDG/McGovern Institute for Brain Research, Beijing Normal University, Beijing, China.

Words of the manuscript text: 1197 words

**\*Correspondence to:** Shuyan Liu, Department of Psychiatry and Psychotherapy, Charité –

Universitätsmedizin Berlin (Campus Charité Mitte), Charitéplatz 1, 10117, Berlin, Germany

Email: siyan908@hotmail.com

24 Improving prognosis and treatment response to substance use disorder (SUD) remains a  
25 challenge. Such an aspiration, however, is confronted by a reality that psychiatry lacks reliable  
26 tests that predict illness risk, treatment response, remission, or recurrence likelihood. Here, we  
27 consider potential advantages in addressing these shortcomings bestowed by the digital  
28 revolution. The overarching challenge is to realise an integration of digital phenotyping and  
29 mobile sensing into SUD diagnostic assessments, coupled to discovery of underlying  
30 mechanisms and delivery of targeted interventions. The attraction of digital personalized  
31 medicine arises out of a deployment capability within a clinical environment, one that addresses  
32 the challenge of scalability.

34 New digital technologies, including digital phenotyping and mobile sensing [1], provide a  
35 platform for ecological momentary assessments (EMA) and interventions (EMI) that have the  
36 capability to assess and change health-related behaviour patterns [2]. These technologies can  
37 capture diverse variables spanning psychomotor activity, mood state, geolocation, and  
38 movement in real time, with a potential to assess specific covert antecedents (e.g., urges and  
39 cravings, lifestyle factors) and environmental determinants (e.g., high-risk situations) of  
40 addictive behaviours. Such metrics, among others, provide inputs for algorithmic analyses to  
41 enable immediate automated risk-evaluation coupled with an evaluation of candidate predictive  
42 models. Once valid predictive models are established, specific and timely interventions can be  
43 triggered to prevent addictive behaviour relapse [2,3].

45 Smartphone-based platforms (e.g., *movisensXS*, *ESm Capture*, and *LifeData*) provide examples  
46 of digital phenotyping tools. While SUDs are marked by multiple physiologic changes (e.g.,  
47 high arousal) in response to cue exposures, mobile sensing tools can capture physiological  
48 indices of cue-elicited craving, for example those that involve a change in physiological (e.g.,  
49 heart rate and skin conductance) reactivity, that can feed into predictive models of relapse [4].

51 Digital technologies have many advantages over conventional laboratory assessments [1,2],  
52 including an enhanced ecological validity consequent upon their deployment in everyday  
53 settings, coupled to an absence requirement for intrusive specialized equipment. Indeed,  
54 targeted intervention programs (e.g., mindfulness meditation and psychoeducation) can be  
55 digitalized and incorporated into smartphone application (app). Furthermore, through use of  
56 analytic approaches (e.g., machine learning) interventions can adaptively tailor to the precise  
57 phase of the disorder and characteristics of at-risk subjects. To improve patient treatment  
58 outcomes., the use of digital technologies can be part of a global patient follow-up scheme [5].  
59 By providing a robust functionality, these tools can aid a reduction in stigma, particularly  
60 embarrassment experiences through help seeking that involve traditional face-to-face  
61 communication interventions [2]. Finally, it does not escape our attention that an added impact  
62 of digital technologies on substance use treatment arises in an age of unforeseen enforced  
63 isolation, quarantine and social distancing [6].

65 We consider an outline realisation of the above-mentioned aims in greater detail. Substance-  
66 associated cues (e.g., alcohol and cigarette advertisements) are known to often elicit craving  
67 and promote substance-seeking responses, particularly in a context of diminished cognitive

68 control [3]. Unique person-specific cues (e.g., the presence of a specific friend or object /  
69 context that is commonly paired with an individual own substance use) encountered on a daily  
70 basis in a manner that enhances momentary substance craving intensity [4]. The latter highlights  
71 that an individual's unique history of substance use provides important predictive inputs for  
72 quantifying a risk of relapse. Indeed, future studies that rely on digital technologies can reap a  
73 rich harvest of data based upon dense sampling of dynamic features of an individual's  
74 experience, environments, or social contexts in real time. In this way digital technologies enable  
75 new opportunities to understand and act upon these person-specific risk factors and individual  
76 lifestyle features, at a level that facilitates individually tailored interventions.

77  
78 Another example of a leverage point for digital technologies is the likelihood of momentary  
79 daily fluctuations in inhibitory control, sensation-seeking, and urgency. Thus, as changes in  
80 working memory function, have been associated with momentary enhancements in craving that  
81 can lead to subsequent SUDs [3,4]. Substance-associated expectancies and availabilities also  
82 mediate cognitive and behavioural impairment in real life [4]. Capturing such fluctuations in  
83 momentary SUD behaviours and experiences across time and in everyday settings hold  
84 potential for informative inputs to predictive models and prevention strategies.

85  
86 Gamified mobile cognitive tests have emerged as powerful tools to advance a mechanistic  
87 interpretability of SUDs. Embedding a standard experimental task, within an engaging, fun-to-  
88 play game is an increasingly popular experimental strategy not least as it improves both  
89 participant engagement and retention. Thus, use of an app enables delivery of a gamified battery  
90 of cognitive tests that measure circadian rhythm and time-of-day variations in cognitive  
91 function in SUDs. A pertinent pioneering example is *The Great Brain Experiment*  
92 (<http://www.thegreatbrainexperiment.com>), which supports a gamified app that transforms  
93 otherwise mundane tasks into an engaging activity. It provides access to cognitive mechanisms  
94 that encompass inhibitory control, working memory, risk taking and Pavlovian bias. Of note all  
95 these processes have potential relevance for understanding SUDs. Another example is a newly  
96 launched *Brain Explorer* app (<https://brainexplorer.net>) which provides a basis for exploring  
97 associations between adolescent cognitive and brain development with a disposition to  
98 substance use.

99  
100 Tracing the trajectory of SUD via EMA and unpicking underlying mechanisms via gamified  
101 cognitive tests, enable researchers to embed EMI into SUD patients' daily lives. EMI can be  
102 delivered in many ways ranging from unstructured recommendations (e.g., requesting a SUD  
103 patient practice relaxation technique when showing an increased stress-induced craving) to  
104 more formalized and structured interventions (e.g., patients participating in a substance use  
105 cessation intervention receive a prompt via an app with tips for dealing with cravings during a  
106 time when they typically use substances). Notably, EMI can supplement existing interventions  
107 including standard ongoing medical and psychological treatments or can be implemented as  
108 stand-alone interventions. Additionally, these technologies take well-developed therapeutic  
109 interventions (e.g., cognitive behavioural therapy and guided self-help) and apply them in a  
110 potentially engaging format as, for example, through exploitation of music and gamified  
111 features. Smartphone-based interventions are not limited to a particular setting and can be

1  
2  
3 112 delivered while a SUD patient sits at home, reducing patient-flow delays and easing human  
4 113 resources burden [7]. An interventional example is a smartphone-based *Addiction –*  
5 114 *Comprehensive Health Enhancement Support System* (A-CHESS, <http://chess.wisc.edu/achess->  
6 115 *archive/*). A-CHESS provides monitoring and support services to SUD patients. Shanghai  
7 116 Mental Health Centre recently developed an app *Community-based Addiction Rehabilitation*  
8 117 *Electronic System* (CAREs), that offers personalized recovery monitoring and enables SUD  
9 118 patients get easier access to social support in community.  
10 119

11 120 There are several implementation barriers in utilisation of digital technologies that deserve  
12 121 mention [1,2,6]. First is a general concern around data protection and privacy. This can be  
13 122 addressed via an anonymization of data and a thorough quality evaluation of apps as to enact  
14 123 the highest data standards and adherence to regulatory requirements. Second is the question as  
15 124 to whether digital tools will be feasible for use and acceptable to SUD patients, as app efficacy  
16 125 depends largely on user engagement. Third is a challenge in building simple, elegant, and  
17 126 intuitive user interface designs especially for patients with low digital literacy. Moreover, many  
18 127 digital tools remain inaccessible to rural underserved areas and vulnerable populations (e.g.,  
19 128 patients who are elderly, living with some form of disability). Finally, the full potential of digital  
20 129 health can only be realised in a world where digital technology is democratically accessible to  
21 130 everyone.  
22 131

### 23 132 **Acknowledgements**

24 133 Funds: Stiftung Charité (Dr Dolan).  
25 134

### 26 135 **References**

- 27 136 1. Insel TR. Digital phenotyping: Technology for a new science of behavior. *JAMA* 2017; 318:  
28 137 1215-1216
- 29 138 2. Baumeister H, Montag C. *Digital Phenotyping and mobile sensing: New developments in*  
30 139 *psychoinformatics studies* Springer; 2019
- 31 140 3. Bertz JW, Epstein DH, Preston KL. Combining ecological momentary assessment with objective,  
32 141 ambulatory measures of behavior and physiology in substance-use research. *Addict Behav* 2018; 83:  
33 142 5-17
- 34 143 4. Heinz A, Kiefer F, Smolka MN et al. Addiction research consortium: Losing and regaining  
35 144 control over drug intake (ReCoDe)—From trajectories to mechanisms and interventions. *Addict Biol*  
36 145 2020; 25: e12866
- 37 146 5. Gründer G, Jungaberle H. The potential role of psychedelic drugs in mental health care of the  
38 147 future. *Pharmacopsychiatry* 2021.
- 39 148 6. Hsu M, Ahern DK, Suzuki J. Digital phenotyping to enhance substance use treatment during the  
40 149 COVID-19 pandemic. *JMIR Ment* 2020; 7: e21814
- 41 150 7. Nesvåg S, McKay JR. Feasibility and effects of digital interventions to support people in recovery  
42 151 from substance use disorders: Systematic review. *J Med Internet Res* 2018; 20: e255  
43 152