



An overview of and recommendations for more accessible digital mental health services

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Abstract | Mental health concerns are common, and various evidence-based interventions for mental health conditions have been developed. However, many people have difficulty accessing appropriate mental health care and this has been exacerbated by the COVID-19 pandemic. Digital mental health services, such as those delivered by mobile phone or web-based platforms, offer the possibility of expanding the reach and accessibility of mental health care. To achieve this goal, digital mental health interventions and plans for their implementation must be designed with the end users in mind. In this Review, we describe the evidence base for digital mental health interventions across various diagnoses and treatment targets. Then, we explain the different formats for digital mental health intervention delivery, and offer considerations for their use across key age groups. We discuss the role that the COVID-19 pandemic has played in emphasizing the value of these interventions, and offer considerations for ensuring equity in access to digital mental health interventions among diverse populations. As healthcare providers continue to embrace the role that technology can play in broadening access to care, the design and implementation of digital mental healthcare solutions must be carefully considered to maximize their effectiveness and accessibility.

Globally, mental health concerns are both common and undertreated¹. In the United States, fewer than half of individuals in need of mental health treatment receive care^{2,3}. Moreover, there are substantial variations in unmet treatment needs by geographic region, with more than 123 million Americans living in federally designated Mental Health Professional Shortage Areas⁴. For individuals who live in areas with mental health clinicians, there remain barriers to accessing mental health care. These barriers include, but are not limited to, stigma, cost, privacy concerns, a preference for dealing with problems independently and time constraints^{5,6}. One study in the United States found that 62% of adults with a mental illness and 41% of adults with a serious mental illness did not receive any mental health care in the past year⁷. For individuals who do seek treatment, many do so in primary care medical settings in which pharmacotherapy is often used with inadequate dosing and follow-up care⁸.

To address these treatment gaps, a range of digital mental health interventions (DMHIs) have been developed over the past 25 years. These interventions have been developed to offer an alternative to traditionally delivered mental health interventions, and may be more accessible and acceptable to some individuals. We use

DMHIs to refer to programmes that deliver psychological strategies and interventions via online and/or mobile platforms. We include telehealth services, in which mental healthcare services are delivered synchronously via a telephone or videoconference-based connection, under the umbrella term of DMHIs.

Many individuals and healthcare providers quickly adopted DMHIs to provide mental healthcare services while maintaining physical distancing protocols during the COVID-19 pandemic^{9,10}. Organizational policies and state laws shifted to accommodate the need to deliver healthcare services at a distance, such as allowing clinicians to deliver remote services from home. Moreover, patients and providers alike became more familiar with the potential benefits of DMHIs¹¹. Thus, the COVID-19 pandemic has hastened the adoption and development of some DMHIs (see BOX 1). Given the widespread utilization of digital tools, it is likely that DMHIs will continue to be used even as physical distancing protocols for infection control are no longer needed. Yet questions remain on how to best integrate DMHIs into future healthcare systems.

More specifically, although many DMHIs show clinically meaningful improvements relative to controls when tested under tightly controlled research settings¹², there is

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a large research to practice gap: many attempts to implement DMHIs in real-world settings have been unsuccessful and people do not engage with them¹³. This is not to say that all implementations of DMHIs have failed. Indeed, numerous Internet-based cognitive behavioural therapy (iCBT) programmes implemented in routine care settings, primarily outside the United States, have produced clinically meaningful effects, albeit to a lesser degree than in tightly controlled trials¹⁴. However, many existing DMHI programmes are not acceptable to the targeted end users and clinicians; in other words, they do not fit into clinicians' workflows and are not aligned with how patients want to use personal technologies¹⁵. This is clearly problematic because if people are not accepting of DMHIs, and thus not accessing these interventions, they are not able to benefit from them. Without acceptability and subsequent adoption of interventions by healthcare providers and patients, efficacy of these interventions is meaningless from a public mental health standpoint.

In this Review, we aim to help inform the next steps for DMHIs that foster the widespread use of digital mental health services. First, we present an overview of DMHIs. We outline the treatment approaches that are used across mental health diagnoses, explain the range of ways in which these interventions are delivered and consider how interventions for individuals in key age groups vary in design, content and delivery methods. Then, we provide recommendations for integrating DMHIs into routine clinical care and promoting the accessibility of these interventions beyond the COVID-19 pandemic, with a particular focus on equity in access among diverse populations.

Therapeutic models

Digital mental health services are typically designed based on psychological treatments that have demonstrated efficacy in traditional, face-to-face mental health care. In this section, we discuss the treatment approaches commonly found in digital mental health services:

cognitive behavioural therapy (CBT), acceptance and commitment therapy (ACT) and psychodynamic therapy (PDT). These therapeutic models were selected for review here due to their common use in face-to-face therapy and the frequency with which they have been used in the development of DMHIs. An overview of the therapeutic rationale and differences between therapeutic models in face-to-face and DMHI contexts is provided in FIG. 1.

We also review the literature on the efficacy and effectiveness of these approaches. Examples of recent meta-analyses on various types of DMHI for a range of treatment targets are provided in TABLE 1. This table is intended to provide a general overview and is not necessarily a comprehensive overview of all DMHI research to date.

CBT, ACT and PDT are not the only frameworks on which DMHIs are based. Other therapeutic models are discussed in BOX 2.

Cognitive behavioural therapy. CBT is a class of psychological interventions based on the premise that cognitive and behavioural factors contribute to the maintenance of emotional distress and behaviour problems¹⁶. For example, individuals might be taught to identify unhelpful thoughts and replace them with more helpful thoughts (a process referred to as cognitive restructuring) and might be prompted to modify maladaptive behavioural patterns. For example, an individual might be taught to identify an unhelpful thought such as 'she did not say hi to me because she's mad at me' and replace it with a thought such as 'she might not have seen me, or maybe she was in a hurry and did not have time to talk'. There are a wide variety of disorder-specific CBT protocols that address specific cognitive and behavioural maintenance factors (for example, biased thinking and avoidance) associated with different mental health disorders (for example, depression and anxiety)^{17,18} and behavioural health issues (for example, pain)¹⁹. CBT sessions with a therapist are typically held once a week for a few months. Meta-analyses of face-to-face CBT for various mental health disorders have found that CBT is either more effective than or comparably effective with other forms of psychotherapy²⁰. Thus, CBT is often considered the gold standard or first-line treatment for numerous psychological disorders²⁰.

CBT is the most researched family of face-to-face psychotherapies. It is also the most researched family of digital mental health programmes. Individual randomized trials and meta-analyses have shown that CBT delivered by telephone (T-CBT) is comparably efficacious with CBT delivered in a face-to-face setting^{21,22}. There is also abundant evidence supporting the efficacy of iCBT, which is often delivered via interactive websites. Indeed, evidence from a meta-analysis and systematic review indicates that iCBT is comparably as effective as face-to-face CBT for depression and anxiety^{22,23}.

There are well-studied commercially available iCBT programmes, such as *Beating the Blues* and *SilverCloud*, that are designed as online courses. In these types of iCBT programmes, users receive access to CBT-based psycho-educational content and are prompted to engage

Box 1 | Role of COVID-19 in expanding the availability of DMHIs

The public health guidance related to the COVID-19 pandemic (for example, keeping physical distance from others and staying home) facilitated an expansion of digital mental health services¹⁷¹. This occurred, in part, because face-to-face services were halted in many areas of the world, and because pandemic-related stress prompted more individuals to seek mental health support¹⁷².

The most notable impact of the pandemic on mental health services was the rapid scale-up of virtual (that is, telephone-delivered and videoconference-delivered) mental health service provision. Studies of this scale-up to virtual mental health appointments indicate that the switch to virtual often took a few weeks¹⁷³. Initially, as appointments switched to virtual, appointment counts decreased, and then increased substantially as time progressed¹⁷³. Furthermore, virtual mental health services attracted a large percentage of patients for whom it was their first time using virtual health services, even in healthcare settings in which virtual services were previously offered^{174,175}.

The COVID-19 pandemic also produced new opportunities for web-based and app-based digital mental health interventions (DMHIs) targeting pandemic-related distress. For example, an individually tailored COVID-19-focused Internet-based cognitive behavioural therapy (iCBT) programme demonstrated preliminary evidence of effectiveness at reducing stress, along with symptoms of depression and anxiety¹⁷⁶. In the United States, a publicly available app targeting COVID-19 distress was developed and attracted nearly 50,000 users in a 5-month period, but clinical outcomes have not yet been published¹⁷⁷.

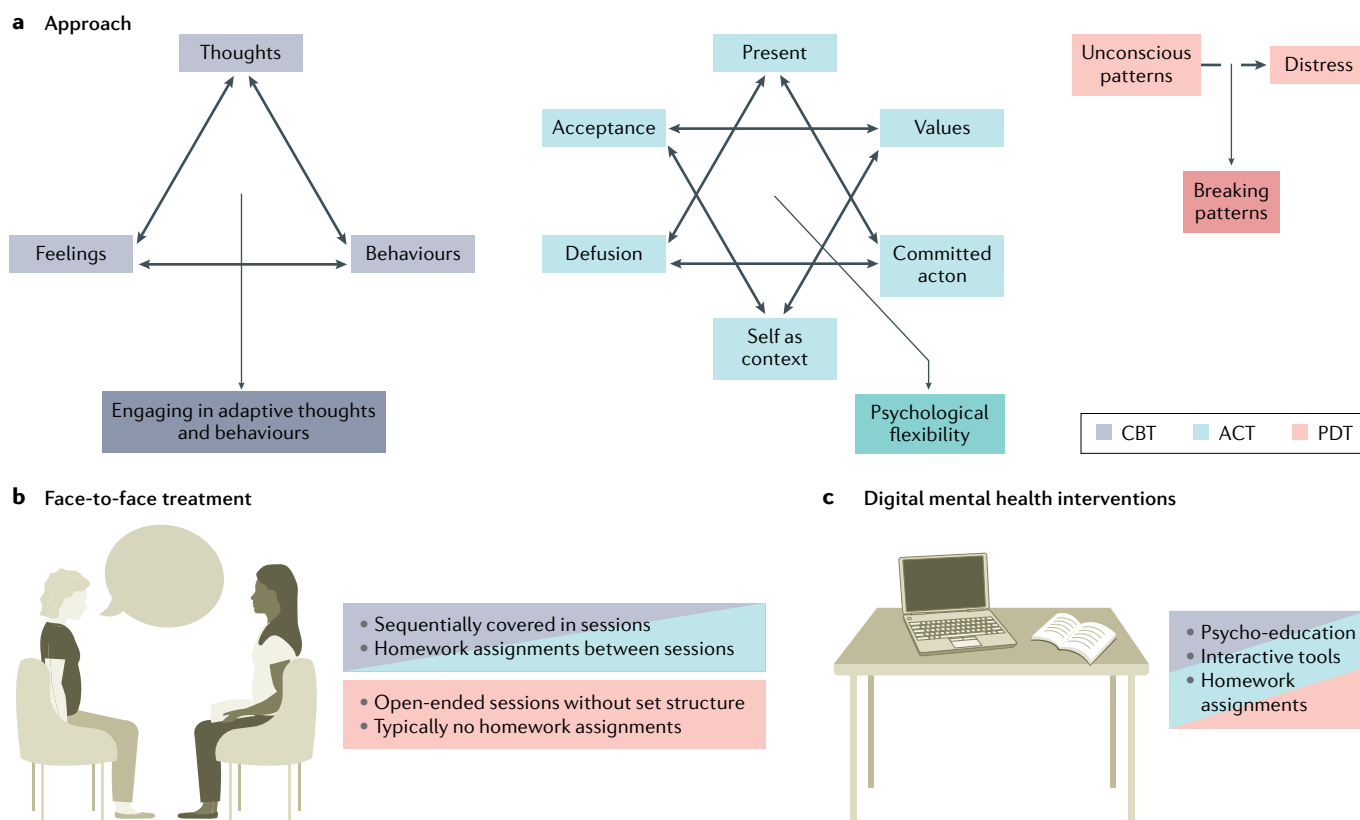


Fig. 1 | Treatment approaches. **a** | The therapeutic rationale underlying cognitive behavioural therapy (CBT; grey), acceptance and commitment therapy (ACT; blue) and psychodynamic therapy (PDT; red). In all of these therapies, individuals are learning more adaptive ways of thinking and behaving, but the means for learning more adaptive skills varies between approaches. **b** | PDT looks very different from CBT and ACT in face-to-face practice. **c** | The psycho-educational content and interactive tools for Internet-based CBT (iCBT) and Internet-based ACT (iACT) are similar. The format of Internet-based PDT (iPDT) also looks fairly similar to both iCBT and iACT in that there are psycho-educational lessons and interactive tools that examine individuals' thoughts and behaviours.

in the same types of exercises or activities that comprise traditional, face-to-face CBT. These iCBT programmes can look and feel similar to an online course, in which users work through lessons or modules and access interactive tools. For example, iCBT programmes typically guide users through cognitive restructuring and use interactive tools to prompt users to modify maladaptive behaviour patterns and engage in more adaptive behaviours. Combining iCBT and T-CBT in stepped care models, in which participants with depression begin with an iCBT programme and are 'stepped up' to T-CBT based on a lack of symptom change, have comparable efficacy with T-CBT alone, with lower overall costs²⁴.

Mobile app-based CBT comes in various forms. Some apps guide users through modules similar to iCBT programmes, whereas others focus more explicitly on the practice of CBT-focused skills such as scheduling positive activities or identifying and challenging unhelpful thoughts. Mobile app-based CBT has a strong commercial presence, with thousands of apps tagged with 'CBT' on the Google Play and iTunes app stores. However, evidence to support the efficacy of CBT-based apps in real-world environments is limited. There are multiple reasons for this limited evidence base.

First, although both iCBT and app-based CBT are typically evaluated in research settings, there are many

more commercially available apps that are accessible to the general public that have not been rigorously evaluated prior to public distribution. Engagement with apps is often suboptimal in real-life settings, perhaps due to the absence of concurrent human support, which is typically made available in a research trial²⁵. Consequently, CBT exposure through app use in real-life settings is low²⁶.

Second, high heterogeneity in methodologies (for example, the outcomes evaluated or whether a comparison group is included)²⁷ and reporting practices for use and efficacy outcomes from app-based CBT intervention trials make it difficult to establish a clear evidence base^{28,29}. One reason for heterogeneity in methodologies is that although randomized trial designs are the most rigorous approach for testing interventions³⁰, such trials can be time-consuming and expensive, and technology evolves at a much faster pace, leading many groups to test effects in smaller, single-arm study designs³¹. However, trial designs and frameworks that encourage optimization without compromising rigour, such as those that evaluate intervention principles rather than specific interventions, are gaining favour^{31–34}.

Third, with noted exceptions — for example, *Woebot*, *IntelliCare*, *SuperBetter* and *WebMAP Mobile* — the majority of CBT apps that demonstrate efficacy in

Table 1 | Example meta-analyses of DMHIs for diverse treatment targets and populations

Article	Number of studies included	Treatment target(s)	Population	Digital mental health intervention	Contrast	Effect size
Andrews et al., 2018 (REF. ⁶³)	53	Depression and anxiety	Adults	iCBT	iCBT vs treatment as usual	$g = 0.38^*$
					iCBT vs waiting list control	$g = 0.90^*$
					iCBT vs face-to-face CBT	$g = 0.14$
Guo et al., 2020 (REF. ¹⁶⁶)	20	Social anxiety disorder (SAD)	Adults	iCBT	iCBT vs control groups	$g = -0.55^*$
					iCBT vs face-to-face CBT	$g = -0.18$
Soh et al., 2020 (REF. ¹⁶⁷)	33	Insomnia	Adults	iCBT	iCBT vs control groups	$d = 0.392^*$
Wootten, 2016 (REF. ¹⁶⁸)	18	Obsessive-compulsive symptoms	Adults	Remote CBT (delivered by computer, telephone and videoconference)	Remote CBT vs control	$g = 1.06^*$
					Remote CBT vs face-to-face CBT	$g = -0.21$
Cervin and Lundgren, 2021 (REF. ¹⁶⁹)	9	Anxiety disorders	Children and adolescents	Technology-delivered CBT (tCBT; delivered by Internet, app, mobile phone or tablet computer)	tCBT vs control	OR = 4.73*
Thompson et al., 2021 (REF. ⁴⁸)	25	Anxiety, depression, quality of life and psychological flexibility	Adults	iACT	iACT vs control groups on anxiety	$g = 0.24^*$
					iACT vs control groups on depression	$g = 0.38^*$
					iACT vs control groups on quality of life	$g = 0.27^*$
					iACT vs control groups on psychological flexibility	$g = 0.32^*$
Lindegaard et al., 2020 (REF. ⁵⁵)	7	Depression, anxiety and quality of life	Adults	iPDT	iPDT vs controls on depression ($g = 0.46$), anxiety ($g = 0.20$) and quality of life ($g = 0.40$)	$g = 0.46^*$
					iPDT vs controls on anxiety	$g = 0.20^*$
					iPDT versus controls on quality of life	$g = 0.40^*$
Linardon et al., 2019 (REF. ⁸⁰)	66	Depression, anxiety, stress, quality of life	Adults	Other: mobile apps	Smartphone apps vs controls on depressive symptoms	$g = 0.28^*$
					Smartphone apps vs controls on generalized anxiety	$g = 0.30^*$
					Smartphone apps vs controls on stress	$g = 0.35^*$
					Smartphone apps vs controls on quality of life	$g = 0.35^*$
Linardon et al., 2020 (REF. ¹⁷⁰)	36	Eating disorders	Adolescents and adults	Other: Internet-based, app-based and CD-ROM-based interventions	Digital interventions vs control conditions in reducing established risk factors and symptoms in prevention-focused trials	g values range from 0.19* to 0.43*
					Digital interventions vs control conditions in reducing established risk factors and symptoms in treatment-focused trials	g values range from 0.29* to 0.69*

Hedge's g effect sizes can be interpreted as 0.2 is a small effect, 0.5 is a medium effect and 0.8 is a large effect. Linardon et al.¹⁷⁰ presented meta-analytic results for 35 outcomes; for brevity, results are summarized in ranges. Cervin and Lundgren¹⁶⁹ presented an odds ratio (OR) rather than a Hedge's g value in their published meta-analyses; the OR indicated that participants who received tCBT were approximately four times as likely as control participants to achieve remission from their primary anxiety disorder. CBT, cognitive behavioural therapy; DMHI, digital mental health intervention; iACT, Internet-based acceptance and commitment therapy; iCBT, Internet-based cognitive behavioural therapy; iPDT, Internet-based psychodynamic therapy. *Statistically significant result.

research settings are not publicly available and only accessible to individuals enrolled in a research study^{35–39}. Thus, data on these interventions in non-research, real-world settings are not available; consequently, effectiveness of their use in everyday, real-world environments cannot be assessed.

Acceptance and commitment therapy. CBT is particularly well researched and prominent in the DMHI arena, but other evidence-based modes of psychotherapy have also been translated into DMHIs. ACT is a psychological intervention framework that focuses on mindfulness and acceptance strategies, along with commitment and

behaviour change strategies, to increase psychological flexibility (defined as being in contact with the present moment and being able to engage in effective, flexible behaviour). There are ACT treatment protocols for a range of treatment targets including, but not limited to, depression⁴⁰, anxiety⁴¹, psychosis⁴², eating disorders⁴³, irritable bowel syndrome⁴⁴ and grief⁴⁵. Within ACT protocols, psychological flexibility is targeted through six processes thought to be relevant across a range of clinical diagnoses and conditions: acceptance (defined as active awareness and embrace of present experience), cognitive defusion (a technique used to help people cope with uncomfortable thoughts and feelings), being present (defined as ongoing, non-judgemental contact with events as they occur), the self as context (the idea that people are not the content of their thoughts and feelings but, rather, are the consciousness experiencing those thoughts and feelings), values (the identification of desired qualities or behaviours that give lift purpose) and committed action (defined as taking action in the direction of one's valued life direction)⁴⁶.

Research studies demonstrate that ACT can be competently delivered by telephone⁴⁷ but there have not been large-scale randomized trials comparing telephone-delivered ACT with face-to-face ACT to determine the effectiveness of telephone-delivered ACT. Other ACT-based DMHIs, including Internet-based ACT (iACT), take the principles of ACT and translate them for an online learning environment. These are often structured similarly to iCBT programmes, in that participants receive access to psycho-educational models and supportive resources and tools (such as mindful audio recordings and reflective journaling entries). A meta-analysis on 25 studies of iACT interventions for

adults that targeted depression, anxiety, quality of life and/or psychological inflexibility found small pooled effects for all outcomes post assessment that were maintained at follow-up time points⁴⁸. Although these results indicate potential for iACT interventions, most of the control groups in these studies were wait-list controls, who receive no treatment. The lack of active controls limits the conclusions that can be drawn because wait-list controls offer a less rigorous test of an intervention's effectiveness⁴⁹. Furthermore, there was limited evidence of reliable, clinically significant effects⁴⁸. Thus, further research is needed to evaluate how iACT interventions perform relative to active controls, who receive an alternative treatment, to determine the comparative effectiveness of iACT and, therefore, which type of treatment should be offered to patients.

Psychodynamic therapy. PDT was developed from traditional psychoanalytic psychotherapy — an intensive psychotherapy in which patients are typically engaged in therapy sessions with a certified psychoanalyst several times a week over many years. PDT is a briefer form of therapy based on psychoanalytic theory of how the mind works. Patients engaged in PDT typically meet with their therapist once or twice a week for a shorter duration of time than in psychoanalytic therapy (such as a few months rather than years). Therapy sessions are focused on exploring aspects of the self that are not fully known, particularly how those aspects are manifested and influenced within the therapeutic relationship. In these sessions, patients might begin to recognize and explore recurring patterns in their lives to see how they avoid distress or develop defence mechanisms to cope; these insights might enable patients to change these patterns.

One review identified the following features as unique to PDT: a focus on affect and expression of emotion; exploration of attempts to avoid distressing thoughts and feelings via specific archetypes (for example, sublimation, a defence mechanism in which unacceptable urges are transformed into more socially acceptable behaviours); identification of recurring themes in the patient's life; discussions of past experiences, with a focus on early life experiences; a focus on interpersonal relationships; a focus on the therapeutic relationship between patient and therapist; and an exploration of the patient's fantasy life in which patients are encouraged to speak freely about whatever is on their minds during sessions, rather than following a more structured format⁵⁰.

PDT has historically been less studied in research trials compared with CBT and ACT. However, a large body of research indicates that patients receiving PDT show greater improvements in symptoms than no-treatment control groups^{51,52}. Other studies have found that patients receiving PDT have similar improvements to patients receiving CBT, although the evidence is somewhat mixed⁵³.

The potential benefits of expanding PDT delivery through remote synchronous models (such as telephone and videoconference) include increased access for individuals who are homebound or do not live near psychodynamic therapists⁵⁴. However, there are no published

Box 2 | Other therapeutic approaches used in DMHIs

Positive psychology interventions aim to increase well-being by cultivating positive feelings (such as gratitude), positive cognitions (such as 'I can handle this') and positive behaviours (such as performing acts of kindness). There is evidence that digital intervention programmes based within a positive psychology framework, which focuses on supporting patients to understand and recognize areas of growth, and work towards resilience and a greater sense of well-being, are effective at reducing depressive symptoms^{38,178}.

Interpersonal therapy is a well-supported treatment for mental health concerns such as depression¹⁷⁹. Interventions focus on improving the quality of a client's interpersonal relationships and social functioning to help reduce their distress. Interventions focus on resolving problems in one of four key areas: interpersonal deficits, unresolved grief, life transitions or interpersonal disputes. Interpersonal therapy has been delivered in digital mental health intervention (DMHI) formats but is less effective than Internet-based cognitive behavioural therapy (iCBT)¹⁸⁰.

DMHIs can also be designed to support couples. There is growing evidence for the efficacy of DMHIs based on integrative behavioural couple therapy^{181,182}. Integrative behavioural couple therapy interventions focus on the emotional reactions of partners to the difficulties they encounter in their relationships. Therapeutic strategies include empathic joining around the problem (in which partners begin to empathize with one another about their problem), unified detachment from the problem (in which partners understand that they have co-created their current situation, and thus the problem) and building tolerance to some of the responses that the problem can trigger (such as becoming less blaming and more supportive while recognizing that the other partner is in pain).

Finally, there are numerous commercially available DMHIs, particularly mobile apps, that are not specifically based within a psychotherapeutic framework¹⁸³. There is no research evidence for the effectiveness of such programmes.

studies comparing remote synchronous PDT with traditionally delivered face-to-face PDT.

Internet-based PDT (iPDT) is structured similarly to iCBT and iACT programmes, with weekly text-based modules, homework assignments and e-mail support from a therapist or coach. iPDT differs from face-to-face PDT in a few ways. First, iPDT includes homework assignments whereas face-to-face PDT typically does not. Second, face-to-face PDT often focuses on a theoretical construct called transference, in which the patient's unconscious conflicts associated with past relationships get triggered within current relationships, including the relationship with the therapist. In face-to-face PDT, issues surrounding transference are typically explored as they emerge, whereas iPDT is largely asynchronous (that is, any communication with a therapist/coach is not in real time). Although unconscious feelings towards the therapist/coach may still arise during iPDT, iPDT interventions are not designed to explicitly address transference.

There is a smaller and weaker evidence base for DMHIs based within a PDT framework relative to DMHIs based in CBT and ACT frameworks. A recent systematic review and meta-analysis of iPDT programmes⁵⁵ found seven randomized control trials of iPDT programmes, which targeted depression^{56–59}, anxiety disorders^{57–61} and work-related stress⁶². Compared with inactive control conditions (which included wait-list controls and unstructured support groups), iPDT produced small improvements in depression, anxiety and quality of life outcomes⁵⁵ and the effect sizes were smaller than those found in meta-analyses of iCBT programmes⁶³. Owing to the small number of studies on iPDT, further research is needed to more conclusively determine its efficacy.

Formats of DMHI delivery

When evidence-based psychotherapies are instantiated in digital platforms, they can be self-guided (for example, the user goes through the intervention independently), delivered with the support of a coach or therapist, or blended with face-to-face treatment.

Guided interventions. Guided interventions are those that include human support as part of their delivery. Guided interventions have been delivered using telephone, video call and/or text message interactions. In studies of guided interventions, support has been delivered by mental health specialists (for example, therapists) as well as non-specialist professionals (that is, paid individuals without healthcare or counselling qualifications). Evidence from randomized controlled non-inferiority trials (that is, trials that test whether one intervention provides at least the same benefit to the patient as an established intervention) that compared technician-assisted iCBT (that is, iCBT delivered by a non-specialist professional) with clinician-assisted iCBT and with a wait-list control group suggests that non-specialist professionals are as effective as mental health specialists in delivering these interventions^{64,65}. This is important because the demand for mental health services worldwide significantly outweighs the number

and capacity of licensed mental health practitioners. Thus, the ability to task-shift the delivery of DMHIs to non-mental health practitioners or lay professionals can enable these tools to be delivered to more people in need.

The role of supporting therapists has been thoroughly examined in studies of iCBT. One study of iCBT therapist messaging found that iCBT therapists frequently encouraged, affirmed and guided their patients, and encouragement, affirmation and self-disclosures from the therapist were correlated with clinical improvement⁶⁶. More recent research has further supported an association between clinical outcomes in iCBT and the content of what supporting therapists write in their messages to users (the primary form of communication in therapist-guided iCBT)⁶⁷.

Meta-analyses have shown that guided interventions have demonstrated efficacy for common mental health problems such as depression⁶⁸, post-traumatic stress disorder⁶⁹, anxiety²³ and insomnia⁷⁰; the presence of a human guide has also been found to increase engagement with interventions in some studies²². Furthermore, meta-analytic reviews indicate that guided iCBT interventions perform as well as face-to-face CBT²². Guided interventions also have demonstrated efficacy for eating disorders and substance misuse, although effect sizes have generally been smaller relative to interventions for depression and anxiety⁷¹.

Because offering human support with a digital intervention requires substantial time and resources, growing attention is being directed to studying conversational agents, or chatbots, to simulate human support⁷². Chatbots use conversational artificial intelligence based on decision trees and advances in machine learning to communicate with users. There are several advantages to using chatbots compared with human support, including reduced reliance on humans, and therefore reduced costs, as well as increased capacity to provide immediate support because chatbots can respond at any time of day, immediately after a user's message. Users have indicated that they feel supported by chatbots even when they know that they are communicating with artificial agents⁷³. However, at present there are risks to these tools. Notably, chatbots can sometimes 'get it wrong' and offer unrelated, unhelpful or inappropriate responses to users' inputs⁷⁴. It is hypothesized that incorrect responses can negatively impact user engagement, and can be especially problematic for addressing users' disclosure of high-risk behaviours of high relevance for DMHIs, such as non-suicidal self-injury or suicidality. Indeed, intervention designers must be attentive to the safety, privacy, liability and ethical considerations associated with using non-human support. It is also unknown how the therapeutic alliance (the relationship between the patient and provider and the degree to which they are engaged in collaborative, purposive work) is impacted by using chatbots⁷⁵. To date, these technologies have not yet achieved capacity to simulate empathy relative to humans⁷⁶. This has important implications for DMHIs because empathy predicts therapeutic alliance and is considered a non-specific factor that can positively impact psychotherapy treatment outcomes^{77–79}.

Self-guided interventions. Self-guided interventions, also referred to as unguided interventions, are interventions that are fully automated and do not involve human support as part of their delivery. With unguided interventions, users can download and access DMHI tools as they desire, and the interventions are designed to be delivered without someone guiding users through the programme. Consequently, unguided interventions have great potential for scale — they can be delivered to large populations of users because the availability of people to serve as guides is not a rate-limiting factor for the number of people who can be provided services. The ability of unguided interventions to function without human support also makes them cheaper to deliver compared with guided interventions. Unguided interventions have demonstrated efficacy relative to control conditions (including wait-list and active controls) for common mental health problems, such as depression and generalized anxiety disorder, but the effects are smaller than for guided interventions that include human support⁸⁰. However, this difference between guided and unguided interventions may be moderated by symptom severity; in a large network meta-analysis of patient data from individuals with depression, the effects of guided interventions relative to unguided interventions were more pronounced for those with higher symptom severity (that is, individuals with moderate to severe depression benefit more from guided interventions)⁶⁸. By contrast, those with mild or subthreshold depression experienced similar effects with guided and unguided interventions⁶⁸.

Numerous commercially available DMHIs offer self-guided and guided versions of their programmes. In this model of contact by choice, users can choose their level of support by opting in and out of contact with a coach or mental health specialist. For example, the chatbot app *Wysa* offers users the ability to exchange messages with a therapist for an additional monthly fee⁸¹. No studies regarding the acceptability and effectiveness of this choice method have been published to date.

Blended approaches. Blended approaches involve delivering DMHIs as part of face-to-face mental health interventions^{82,83}. For example, a digital intervention could be used to facilitate self-monitoring (for example, symptom assessments) to guide treatment decisions, support skill practice between sessions (for example, as an alternative to worksheets) and/or enable additional communication streams between the therapist and the patient (for example, text messaging). Therapists have favourable perceptions of blended treatments, and the use of blended treatments is increasing⁸⁴. In 2014, surveys revealed that mental health service stakeholders were more interested in implementing blended treatment than stand-alone Internet-based interventions⁸⁵; a subsequent survey of therapists in 2019 showed that blended treatments were perceived to have fewer disadvantages than stand-alone Internet-based interventions⁸⁴. Trials testing blended approaches have been conducted. For example, two trials have demonstrated the efficacy of iCBT combined with synchronous chat with a therapist compared with a minimal-attention control (in which participants completed weekly mood assessments

that were reviewed by a therapist) for depression in adolescents^{86,87}. However, among people with psychosis, blended treatment for paranoia (comprising face-to-face sessions, a mobile app and usual care) did not outperform usual care alone (which typically involved antipsychotic therapy, contact with a mental health worker and outpatient psychiatric appointments)⁸⁸.

Regarding the cost of care, a naturalistic study (in which participants received treatment through their standard healthcare setting) found that iCBT blended with face-to-face treatment was more costly to deliver and did not necessarily offer greater benefit compared with face-to-face treatment alone for people with anxiety or depression⁸⁹. Further, many practising clinicians report using mobile apps with their patients in standard practice, but concerns about costs to the patient, privacy and data security are barriers to their use^{90–92}. Continued research is needed to understand why and for whom blended care works. To this end, there are efforts to create instruments that help clinicians identify patients most appropriate for this treatment delivery approach^{83,93}.

Considerations across the lifespan

DMHIs that are marketed for general audiences have typically been evaluated and/or developed with adults who are young (aged 18–40 years) or middle-aged (aged 40–60 or 65 years). However, multiple interventions are targeted towards specific age groups, such as youth (a term that may encompass children, adolescents and young adults up to age 25 years) or older adults (typically defined as those older than age 60 or 65 years). It is important to understand the design and effectiveness of DMHIs for youth and for older adults as these two groups have developmentally different needs and preferences.

DMHIs for youth. DMHIs for youth have been heavily focused on computer (for example, CD-ROM), web (for example, secure Internet page) and mobile-based (for example, text messages or app) delivery^{94–96}. For computer and web-based DMHIs, CBT has been the most commonly used treatment approach^{94,97}. In children 6–18 years of age with depression and/or anxiety, iCBT has yielded medium intervention effects compared with wait-list control groups, but has not demonstrated a statistically significant benefit compared with face-to-face delivery of CBT⁹⁷. However, iCBT programmes evaluated in published studies often rely heavily on psycho-education via text-based lessons, which is unlikely to be engaging to youth⁹⁸.

Computer and web-based DMHIs for youth have also been based on other psychotherapies (for example, problem-solving therapy and attention bias modification training), with little to no effect compared with control conditions⁹⁷. A small, but growing, body of evidence supports the efficacy of mobile DMHIs in improving paediatric health targets (for example, physical activity promotion or medication adherence)⁹⁹. Less support has emerged for the efficacy of mobile DMHIs for mental health targets in youth¹⁰⁰. Nevertheless, the mobile DMHI marketplace is continually growing^{101–103}, raising

concerns about the small evidence base for mobile DMHIs for youth (particularly children) compared with the number of commercially available apps^{96,97,104}. Indeed, some mobile DMHIs are marketed to youth but have only been evaluated with adults. For example, Woebot, an app that uses a CBT-based conversational agent, is marketed for age 12 years and older. However, to date, this app has only demonstrated efficacy in decreasing depressive and anxiety symptoms in college students, and some evidence for efficacy in improving substance misuse behaviours and cravings in adults^{36,37}. Frameworks that clinicians and caregivers can use to identify digital mental health tools for youth needs are being proposed to supplement the gap between established efficacy with adult populations and availability to youth¹⁰⁵.

Youth face numerous barriers to traditionally delivered mental health services, including stigma, their family's negative beliefs about mental health services and low mental health literacy¹⁰⁶. For youth, DMHIs can be solely youth-facing³⁶ or involve shared access with a caregiver¹⁰⁷. Caregivers (often parents) are sometimes included as end users of DMHIs for youth due to the developmental needs of younger users who may require assistance with logistical aspects of an intervention, and findings that intervention outcomes improve when caregivers are involved^{97,99}. Notably, despite this association between caregiver involvement and outcomes, parental consent to mental health treatment can also be a potential barrier to youth mental health treatment¹⁰⁸. Thus, the effect of caregiver involvement in intervention efficacy might be dependent on the specific context, and DMHIs could offer a particularly important mental health option for youth who might otherwise not seek care due to barriers, including the need for parental consent¹⁰⁹. DMHI outcomes for youth also improve with other adult support, such as from therapists and teachers⁹⁷.

Design needs^{110,111} (for example, providing flexibility and an age-appropriate cognitive load) and frameworks specific to youth (for example, the Integrate, Design, Assess, Share Framework¹¹²) have been described elsewhere. Other broad design considerations for youth DMHIs include attention to language (for example, avoiding jargon and anticipating variability in health and general literacy)^{113,114}; brevity (for example, youth often have a lower ability to maintain and manipulate information in their minds than do adults, so scrolling through text, particularly on smaller screens, can be challenging)¹¹⁵; and clear confidentiality and data sharing practices to increase feelings of safety about disclosing information in online spaces, particularly for youth from vulnerable populations^{116,117}.

DMHIs for older adults. DMHIs for older adults are associated with a large, strong research base. Interventions with an evidence base for older adults have primarily focused on computer and web-based interventions that rely heavily on psycho-education and skill-building, and often include some form of human support^{118,119}. For example, therapist-guided iCBT programmes for older adults with depression and anxiety have demonstrated both efficacy (with large clinical improvements in

depression or anxiety) and cost-effectiveness^{119,120}. One study found equivalent benefits for anxiety symptoms from a blended form of ACT, which included a web-based intervention and four face-to-face sessions, and a traditional face-to-face, four-session CBT intervention; however, the blended ACT programme was associated with higher levels of patient satisfaction and well-being¹²¹. This suggests that interventions that include a combination of face-to-face and digital services may be of particular value for older adults.

It is important to note that older adults may not be as comfortable with technology, and some older adults view digital innovations as exclusionary rather than stimulating^{122,123}. The digital divide between generations has been narrowing^{124,125}, yet older adults might still lack experience with and access to digital tools, and might therefore face barriers to the effective use of technologies¹²⁶. Indeed, many mobile health apps are not appropriately designed for and accessible to older adults, who often have low technical readiness and trust in technologies^{127,128}. Although DMHIs for older adults might overcome some common barriers to traditionally delivered mental health services, such as cost, stigma and transportation issues, implementations of DMHIs for older adults should strategically target trust, buy-in and technical skill training or support to maximize the likelihood of success.

Towards greater and more equitable access

The more widespread use of DMHIs due to the COVID-19 pandemic is itself a research opportunity — as more people are using these services across healthcare settings, there are rapidly accumulating data on the acceptability and effectiveness of these services. As these data are analysed, we will be able to identify the barriers and facilitators to effective DMHI delivery. Early data generated during the COVID-19 pandemic has demonstrated that the rapid scale-up of telemedicine in the United States decreased mental service use by patients older than age 65 years, those who prefer a language other than English, and patients on Medicare and Medicaid (federal health insurance programmes available to individuals based on income, age and disability status). Furthermore, patient health visits for white patients increased, whereas visits decreased among other racial and ethnic groups¹²⁹. Thus, the initial scale-up of telemedicine likely increased healthcare disparities for more vulnerable patient populations. It is clear that we need to do better. In this section, we outline opportunities to create more equitable access to mental health care. Due to the nuances of payment systems and social structures within the American healthcare system, this section focuses primarily on data from the United States. A brief global perspective on access is provided in BOX 3.

Efforts towards greater and more equitable access to mental health care must be strategic in continued expansions of DMHIs. In the American healthcare system, in which most patients are on private health insurance plans, issues of payment and cost are front and centre. To ensure access to mental health care, it must be both available and affordable. The COVID-19 pandemic facilitated a rapid transition to remote care by relaxing

Box 3 | Examples of access to DMHIs around the world

High-income countries

In high-income countries, digital mental health interventions (DMHIs) can improve access to mental health care and are being offered in both self-guided and guided formats. In addition, some countries have integrated DMHIs into their public health insurance models. For example, in Germany the Digital Healthcare Act entitles all individuals covered by public health insurance to be reimbursed for their use of certain digital health applications. There is now a regulatory process to approve specific DMHIs for reimbursement¹⁸⁴. In the United Kingdom, the Improving Access to Psychological Therapies programme through the National Health Service (NHS) provides computerized cognitive behavioural therapy (CBT) and Internet-based cognitive behavioural therapy (iCBT) free of charge as a low-intensity treatment for depression and anxiety disorders¹⁸⁵. The NHS offers an [online library](#) of mental health apps to guide patients to the appropriate apps. Finally, in Australia the Department of Health developed the [Head to Health](#) platform, which serves as a gateway to mental health resources, including DMHIs. Several DMHIs are available for free to Australians with a prescription and for a small fee without a prescription¹⁸⁶.

Low and middle-income countries

In low and middle-income countries, stand-alone, scalable DMHIs could have a large impact on population mental health given the scarcity of mental health professionals in many countries¹⁸⁷.

A recent review of DMHIs in low and middle-income countries identified 37 studies, with about half ($N=23$) reporting clinical benefits to participants¹⁸⁸. Sample sizes were small, and studies primarily focused on adults. Studies were primarily from East Asia, Central or Latin America and South Asia¹⁸⁸. There remains a need to identify the most effective DMHIs for these resource-limited settings and to integrate high-quality DMHIs into health systems in these areas.

rules for telehealth services (which were reimbursable under certain insurance plans only to individuals who lived in specific geographic areas or had specific diagnoses), and many states adopted telehealth parity through COVID-19 emergency orders. Telehealth parity guaranteed that clinicians would receive comparable payments for telehealth as for in-person services¹³⁰. To integrate DMHI services into routine clinical practice and ensure that they continue to be available beyond the COVID-19 pandemic, telehealth parity laws should continue to expand and be made permanent.

There are numerous healthcare payment models within the United States, and many of these payment models create challenges for reimbursement for DMHIs other than synchronous telephone or videoconference appointments with clinicians. Organizations that operate outside fee-for-service models may be well positioned to offer DMHIs as part of their new standard of care. However, the majority of healthcare payments in the United States are tied to the fee-for-service model of health care; thus, care is not accessible to those with limited payment options. There is an urgent need to expand reimbursement for DMHI products, such as for mobile apps and web-based programmes, that can be provided with or without human support and for a wider range of clinicians who could support DMHIs¹³¹.

Much of the COVID-19 scale-up of telemedicine relied on videoconferencing in lieu of in-person appointments. Videoconferencing appears to be an effective and engaging method of delivering mental health care^{132,133}. However, when videoconferencing is the only service available, patients can be left behind if they lack access to or comfort with technology capable of videoconferencing¹³⁴. Using mobile phones as a delivery

method might better engage vulnerable patient populations compared with computers or other technologies that are reliant on Wi-Fi to access DMHIs¹³⁵. Indeed, there remain disparities in who is likely to adopt computer use and have home broadband access¹³⁶. However, nearly all adults (96%) in the United States own a mobile phone (that is, a cellular phone that supports telephone calls and/or SMS messaging)¹³⁷ and 95% of teenagers report that they either own or have access to a smartphone (that is, a cellular phone that can access the Internet and apps)¹³⁸.

The bulk of evidence for DMHIs has focused on computer-based platforms. However, mobile phone-based interventions (and, specifically, interventions that do not require smartphone ownership and Wi-Fi access) are more likely to reach vulnerable patient populations. Moreover, despite the huge number of mental health apps, adoption is often poor²⁶. SMS or text messaging-based interventions offer numerous potential benefits for expanding DMHIs and making DMHIs more accessible. First, text messages are more likely to be read than push notifications, which are often easy to ignore¹³⁹. It is theorized that users might prefer text messages rather than having a potentially identifiable mobile app saved on their smartphone (for example, a specific app could identify the user as having depressive or anxiety symptoms)¹⁴⁰. Second, academic clinical scientists are unlikely to develop, evaluate and disseminate apps with the same level of speed as found with industry-driven apps, but can often rapidly program and disseminate text message-based DMHIs¹⁴¹. These differences in end-user engagement and resources make the development of future apps targeted specifically for unique conditions and/or patient groups less feasible without industry support, whereas text message-based programmes may be more feasible. However, hidden costs of text messaging-based interventions need to be considered, as vulnerable patient populations are more likely than affluent individuals to be impacted by lapses in service or burdened by the cost of data packages or texting plans^{142,143}.

Regardless of the technology used to deliver DMHIs, there is a clear need to involve representative end users and stakeholders in the design and development of these interventions to ensure that research samples are representative¹⁴⁴. This is particularly important to consider when working towards more equitable access, as white individuals, and particularly white young to middle-age adult women, are typically over-represented in DMHI research trials^{145,146}. To develop programmes that will be engaging for more diverse patient populations, stakeholders and end users from those populations need to be included in both the design and testing of the programme. Including both co-design elements and representative participation in clinical trials can enable early identification and remediation of problems with DMHIs for minoritized individuals.

Furthermore, DMHI programmes need to be developed based on both evidence-based treatments as well as theories that recognize the intersectional identities of members of underserved populations. Including a minority stress theory-based perspective is important.

According to the minority stress model, having a minoritized status leads to increased exposure to stressors and this increased stress can lead to poor health outcomes¹⁴⁷. For example, the HORIZONS intervention, which targeted sexual risk behaviours in African American adolescent females, incorporated both an evidence-based treatment^{148,149} and minority-based theory, and demonstrated effectiveness for reducing targeted risk behaviours^{150,151}. Developing partnerships with public and private organizations serving traditionally underserved populations is key for leveraging the complementary strengths needed to design and deploy inclusive, culturally responsive and innovative mental health programmes¹⁵².

To best serve diverse populations of users, DMHIs need to be available in various languages. Adapting evidence-based DMHIs into a new language typically requires cultural adaptations, rather than a straight language translation. For example, in the adaptation of an iCBT programme for Colombian university students,

the ‘Space from Depression’ programme was renamed ‘I can feel better’. This new name was considered a culturally suitable descriptor of improving mood that avoids the cultural stigma related to the label of ‘depression’¹⁵³. When the Australian Mental Health eClinic was redesigned and adapted for Spanish speakers, participants identified the need for more tailored assessment tools, and greater integration with other Spanish language resources and communities¹⁵⁴. Owing to differences in the length of written languages (for example, Spanish and French texts are approximately 20–30% longer than English texts), interface changes often need to be made to ensure that the layout of web-based and app-based programmes is appropriate and usable.

Finally, as discussed above, providing support in an automated manner or directly from a human supporter is valuable for engagement and clinical outcomes^{68–70}. To make DMHIs more accessible and equitable, human support that is representative and promotes diversity must be included. Consistent with recommendations from the US Department of Health and Human Services’ Office of Minority Health, one way to improve mental health disparities is to increase clinical workforce diversity¹⁵⁵. Thus, human support staff for DMHIs should include individuals with demographic membership that is in some way representative of the patients being served by the DMHIs. To move the field forward in promoting greater access to digital mental health care, we propose adopting the recommendations originally presented by Galán et al.¹⁵⁶ and adapted for a DMHI context (see BOX 4).

Box 4 | Applying a racial justice lens to DMHI research

The following recommendations, originally presented by Galán et al.¹⁵⁶ and adapted to digital mental health intervention (DMHI) contexts here, should be adopted to move the field forward in promoting greater access to digital mental health care.

Involve diverse community stakeholders

To combat long-standing racial injustices, researchers must engage members of underserved and minoritized communities in all stages of the research process and ensure they are benefitting from the work in a direct way.

For example, collaborators representing various backgrounds, including community members, should be involved in developing DMHIs and selecting the intervention targets. Who the DMHI is serving, how previous research has considered (or not considered) race and how that influences the questions being asked in the current research or content provided in the DMHI should be considered.

If research is being conducted in a hospital, in an academic institution or through a private company, collaborations with community members can help to address valid mistrust of these institutions among minority individuals.

Stakeholders from the community should also be involved to ensure that study findings are shared with the community (for example, through social media, infographics or newsletters).

Use inclusive research designs and recruitment procedures

Randomized controlled trials often use samples in which people with minoritized identities are under-represented¹⁸⁹, which can limit the public health value of DMHI research¹⁴¹. Research frameworks and designs that mitigate selection bias and target inclusion of minoritized individuals and underserved communities should be considered.

For example, deploying interventions on technologies that participants readily have access to, meeting participants in preferred locations, such as local libraries or their homes, and outside standard work hours, and providing access to food during or after participation can encourage people with transportation barriers, who work full time or who lack private spaces to participate¹⁸⁹. Compensation for participation in DMHI research should adequately reflect the investment of time and disruption to daily life (for example, childcare or transportation costs should be compensated).

The samples with whom measures or assessment tools were validated and how this impacts their use in other groups should be considered. For example, researchers should ensure that the scales used elicit valid and effective responses for all racial groups (for example, Likert scales have been shown to be culturally insensitive or meaningless to some racial groups)¹⁹⁰. In addition to assessing deficits, strengths-based measures, such as psychological well-being and flourishing, should be included to detect the full effects of interventions.

Finally, the research team of individuals who will be interacting with participants should be diverse and inclusive to maximize the likelihood that a diverse range of individuals will participate.

Summary and future directions

Personal technologies, such as computers, tablets and mobile phones, continue to be ubiquitous. Thus, the potential roles of these technologies in supporting mental health and wellness are numerous and important. Here, we have described numerous DMHIs capable of producing meaningful changes in the lives of individuals struggling with various mental and behavioural health concerns^{48,63}. We have highlighted that the evidence regarding clinical effectiveness is robust for remote, synchronous versions of evidence-based mental health treatments (for example, delivered by telephone or videoconference)^{21,157}. Similarly, the evidence of treatment effectiveness is robust for many web-based interventions, such as iCBT^{68,70}, and is growing for app-based and SMS-based interventions^{140,158}. Many of these interventions produce larger effects when paired with some level of human support or guidance⁶⁸, yet self-guided interventions can be effective and are valuable given their scalability⁶⁸. However, there remain challenges in the field of digital mental health, and therefore areas where continued progress is needed.

First, future research should include representative populations and engage with appropriate stakeholders outside academia. Although numerous DMHIs have demonstrated efficacy in randomized controlled trials^{22,97}, minoritized individuals are typically under-represented in these trials^{145,146}. There are concerns that DMHIs believed to be efficacious may not be appropriate for or helpful to individuals from

underserved communities¹⁵⁹. To fully realize the potential of DMHIs, these interventions must be capable of eliminating, or at least narrowing, disparities in access to mental health care. Thus, to be confident in the effectiveness of DMHIs, they should be subject to further testing with diverse populations in real-world environments. Industry–academic collaborations might be particularly effective to this end because industry partners have access to design and development resources, and academic partners complement these resources by contributing their skills in psychological theory, research design and analysis.

Second, there are substantial areas for improvement regarding data privacy. There are legitimate concerns about privacy policies and problematic data sharing within digital mental health platforms that have limited their adoption^{160,161}. Industry–academic partnerships must be thoughtfully built to maximize privacy and trust from potential DMHI end users. Some people trust universities as benevolent overseers of research data because university-sponsored research is subject to an ethics review. However, those who are less inclined to trust university systems because of past or ongoing wrongdoings might be more likely to engage with industry-focused projects and programmes.

Reviews of mobile mental health apps have found that the majority of commercially available apps do not include a privacy policy or terms of agreement, the reading level of existing written policies is frequently inaccessible to the general population and many of the existing privacy policies state that users' information may be shared with third parties¹⁶². Thus, there are clear ethical concerns in the selection of digital mental health tools. To support ethical decision-making, Nebeker et al.¹⁶³ developed a decision-making tool called the Digital Health Checklist. This checklist walks evaluators through four intersecting domains: access and usability; risks and benefits; privacy; and data management. These domains were identified as the key factors critical to decision-making, and can be useful for technology developers, ethics boards, clinical personnel and people considering participation in a digital health study¹⁶³.

To ensure ethical application of DMHIs, we recommend broadscale adoption of the Digital Health Checklist to guide DMHI selection.

Third, many DMHIs engage users in a structured manner. For example, many web-based programmes are module-based, in which users progress through modules based on time-based rules or completion-based rules. Videoconferencing programmes are synchronous, and sessions typically take place during pre-scheduled times. Many app-based programmes have tools that are continuously available to users, and users can access them at any time. However, DMHIs that are more responsive to the users' context are being developed, such as just-in-time adaptive interventions¹⁶⁴. Just-in-time adaptive interventions deliver intervention components to individuals in the right moments to support behaviour change. These interventions work by leveraging smartphone technologies such as sensing to deliver relevant in-context interventions in real time, and offer an increased likelihood that a user will receive the right intervention in the right place and time¹⁶⁴. Micro-randomized trial designs (in which intervention options are randomly assigned at each relevant decision point to enable researchers to examine the causal effect of different intervention options) can be used to test just-in-time adaptive interventions and inform their optimization¹⁶⁵.

Finally, continued expansion of DMHIs may benefit from increasing the automaticity of these tools. As we have indicated, DMHIs that include human support are often more effective than self-guided or automated treatments⁶⁸. However, reliance on human support may limit the broad dissemination of these interventions because there are limits on how many individuals any one person can serve. It is therefore important for the field of digital mental health to continue exploring ways to encourage user engagement and support skill practice via DMHIs without human support. Continued progress in these areas will be of great value for increasing the scalability of DMHIs.

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The authors contributed equally to all aspects of the article.

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