Self-Efficacy, Internet Self-Efficacy, and Proxy Efficacy as Predictors of the Use of Digital Social and Health Care Services Among Mental Health Service Users in Finland: A Cross-Sectional Study

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Background: The COVID-19 pandemic has accelerated the digitalization of services, but at the same time, some user groups are excluded due to a lack of digital skills. One of the vulnerable groups is people with mental health problems. It has previously been found that self-efficacy beliefs explain the use of digital technology, but the role of social support is largely unexamined. Bandura's concept of proxy agency provides a useful concept to investigate the role of social support in using digital services.

Objective: The study investigates sources of social support and the effect of proxy agency, Internet efficacy, and self-efficacy beliefs on the intention to use digital health and social care services.

Methods: Participants (N = 121) were users in community-based mental health services who were recruited through four organizations in southern Finland for a cross-sectional study. The scales for proxy agency and intention to use services were developed for purpose of the study. The questionnaire also comprised self-efficacy and Internet efficacy scales. Quantitative data were analyzed using descriptive statistics and linear regression analysis.

Results: Both Internet self-efficacy and proxy efficacy ($p < 0.001$) predicted the intention to use digital health and social care services. Self-efficacy had an indirect positive effect on intention through Internet self-efficacy. Family members and friends were the main source of social support in using digital services.

Conclusion: Internet self-efficacy and proxy efficacy are important contributors to the intention to use digital health and social care services. Proxy agents may serve as facilitators to reduce the digital divide and promote the use of digital services by complementing individuals' self-efficacy and Internet self-efficacy beliefs. It is important to develop different forms of social support, such as peer support, to enhance the quality of proxy agency. The limitations of proxy agency are discussed.

Keywords: digital services, Internet self-efficacy, mental health service users, proxy efficacy, self-efficacy, social support

Introduction

Background

In the field of social and health care, new digital services are being offered to clients at a growing rate. These services allow and require clients to play a more active role in caring for their health and well-being. The COVID-19 pandemic has further increased the launch of digital services. The focus of care delivery has shifted towards digital alternatives over traditional face-to-face appointments whenever feasible. Rapid changes require adaptation from organizations, professionals, and customers alike. However, the proliferation of digital social and health care services may accelerate digital and social exclusion and increase inequality unless the most vulnerable groups are also able to access services.^{1,2}

One vulnerable group is people with mental health difficulties who would benefit from digital services and who are also willing to use them.^{3,4} For example, a meta-analysis of 13 studies comparing Internet-delivered therapy and face-to-face therapy indicated that treating depression and anxiety disorders with online therapy was as effective as meeting in person.⁵

Both material deprivation and inadequate skills are contributing factors to digital exclusion,^{1,6,7} and rapid advances in digital technology increase the barriers to digital engagement even further.¹ Affordability issues may prevent people with mental health difficulties from having access to the Internet.^{8,9} Furthermore, the opportunity to use the Internet itself does not necessarily secure access to web-based mental health resources. Users may be unaware of services provided in digital form¹⁰ or may experience difficulties due to outdated and incorrect information in the absence of reliable curation.¹¹ Mental health problems themselves have an impact on digital exclusion when symptoms cause difficulties in using digital services and negatively affect one's feelings of competence and self-efficacy beliefs.⁷ In conclusion, mental health service users who might greatly benefit from web-based interventions are not able to access them.⁹

This study examines how self-efficacy, Internet self-efficacy, and proxy efficacy are associated with the intention to use digital social and health care services within a group of adult mental health service users. The reported sources of support to use digital services are also explored. The main theoretical starting points are related to the concepts of self-efficacy and proxy efficacy beliefs^{12,13} as well as the need and quality perspective of proxy agency.^{14,15} Groups with mental health problems benefit from and are willing to use digital social and health services.³ The prerequisite is to ensure that they have the basic skills needed to use them. It is essential to identify the lack of digital skills and the consequent risk of exclusion. Some studies have investigated the impact of different forms of efficacy beliefs and found that Internet self-efficacy and proxy efficacy promoted the use of services.^{16–18} However, the research on proxy efficacy beliefs related to digital social and health care services, especially in the mental health context, is lacking, and further knowledge is required to secure the availability of adequate and high-quality support sources for people with mental health difficulties. Mental illnesses can cause a variety of symptoms that may hinder the use of digital services.⁷ Therefore, it is paramount to identify the different roles of efficacy beliefs and the sources of obtained support and their effect on the intention to use services. The types of support related to the use of digital services provide new information on what should be developed and at what scale.

Self-Efficacy

Albert Bandura's^{13,19} socio-cognitive approach provides an established psychological model of human agency, which is explicated as individual or personal agency. Bandura argued that the most pervasive determinant of personal agency is self-efficacy, ie, people's beliefs in their capacity to produce desired results and forestall detrimental ones by their actions. Self-efficacy beliefs promoting optimism and self-enhancement play a pivotal role in the self-regulation of motivation through goals, challenges, and outcome expectations. There is evidence that self-efficacy beliefs can affect perceived usefulness and perceived ease of use of technology in general²⁰ and especially health informatics applications.²¹

Furthermore, previous research suggests that an assessment of self-efficacy beliefs should correspond to the targeted performance and domain of interest.²² Thus, the construct of domain-specific self-efficacy should have stronger predictive power than the general self-efficacy scale.²³ In health care contexts, for example, application-specific self-efficacy has been defined as an individual's judgment about their own ability to use health information portals (HIP) to acquire health information and services.²⁴ It has been suggested that general self-efficacy would affect an individual's task-specific self-efficacy within the computer domain²⁵ and even have a cumulative effect on software-specific self-efficacy.²⁶ The suggested links between general self-efficacy and task-specific self-efficacy, however, have not been empirically established.

Previous studies show that Internet self-efficacy is positively related to Internet usage¹⁸ and that individuals' confidence in their own abilities to effectively master the new technology is important for adopting it.²⁰ Application-specific self-efficacy also predicts consumer acceptance of health information portals.^{21,24} The technology acceptance model (TAM) is a widely used theory to explain an individual's behavioral intention to adopt a particular piece of technology. The model approaches the acceptance of technology by examining the individual's attitude from the viewpoints of perceived usefulness and perceived ease of use.²⁷ Internet self-efficacy can be considered as an antecedent to perceived ease of use. There is evidence that health information technology (HIT) self-efficacy predicts behavioral intention to use digital services.²⁸

Proxy Efficacy

Bandura¹³ points out that personal agency is not sufficient to explain all forms of human agency because many things individuals seek can only be achieved through interdependent effort. He introduces the concept of proxy agency, in which

people try by one means or another to get those who have access to resources or expertise, or who wield influence and power to act at their behest to secure outcomes they desire. (p. 13)¹³

In some circumstances, individuals lack direct control over social conditions and institutional practices. Through proxy agency, individuals pursue well-being, security, and valued outcomes by trying to get those who have the expertise or influence and the power to act for them.¹³ For example, in the context of health care, patients' confidence in professionals' competence is termed proxy efficacy, which reflects an individual's confidence in the skills of a third party to act effectively on their behalf.²⁹

Previous studies indicate that social support is also a fundamental part in adopting Internet use and new digital technologies.^{16,30} However social support has so far not been officially conceptualized as proxy agency. There is evidence that support from friends and family can encourage Internet adoption, the continuance of Internet use,³¹ and better Internet skills.³² Studies among older adults show that they rely on relatives and friends, rather than professionals, in seeking help and support for heterogeneous information and communications technology (ICT) needs.^{30,33} Social support from close people includes both informational and emotional support,^{17,34} provides possibilities for observational learning, and encourages the use of ICT technologies.¹⁶ Close people are favored over professional helpers, due to their in-depth knowledge of the novice user.^{30,33}

Bandura¹² further states that proxy efficacy beliefs influence self-efficacy, and both types of efficacies have significant roles in behavioral change. Accordingly, stronger perceptions of behavioral self-efficacy and trust in health care providers' capabilities have been found to predict better adherence to health behavior change and promoting health.^{35,36} In particular, proxy agency has been intensively studied within cardiac rehabilitation settings. Studies have shown that proxy efficacy beliefs and self-efficacy on exercise intention are positively associated,³⁷ and proxy efficacy beliefs depend heavily on the participant's experiences of the proxy agent.^{15,29} Furthermore, strong positive proxy efficacy beliefs, as well as self-efficacy beliefs, are evident when the proxy agent is supportive and engaging. Nonetheless, the amount of time spent with the proxy agent does not seem to correlate with proxy efficacy.¹⁵

In an ideal situation, proxy agency is a valuable vehicle or enabler to promote self-development. It helps individuals survive in diverse environments, find ways to overcome various physical or environmental limitations, and model behaviors to achieve their goals. However, proxy agency may carry some risks and causes for concern. Proxy control can serve as a means to avoid taking responsibility for one's actions and saving oneself from laborious tasks. An individual's personal competencies may not be developed, and there is vulnerability in relying on the competence, power, and favors of others.¹³ There is evidence that the convenience of proxy Internet use may inhibit the acquisition of Internet skills,^{38,39} and over-reliance on proxy may also endanger the development of personal efficacy beliefs.¹⁴ Hänninen et al³⁰ found that older adults' digital skills were varied, and some remain dependent on the continuous assistance of extended family members.

Prior studies have also established links between Internet skills and several demographic factors, including gender, age, and educational level.^{40–42} Thus, it is justified to analyze the effects of background factors, age, education level, and relationship status, along with self-efficacy, Internet self-efficacy, and proxy efficacy on the intention to use digital social and health care services.

Aims of the Study

The first aim of this study was to examine the relationship between proxy efficacy, general self-efficacy, Internet self-efficacy, and the use of digital social and healthcare services among adult mental health service users. The second aim is to gain further understanding of social support in using digital services among mental health service users. This study is warranted because previous research has not considered general self-efficacy, Internet self-efficacy, and proxy efficacy together as explanatory variables for the intention to use digital services. To serve the aforementioned aims, the study addresses the following research questions:

RQ1 Do proxy efficacy, general self-efficacy, and Internet self-efficacy predict the intention to use digital social and health care services among mental health service users?

RQ2 Who do mental health service users reach out to for support using the digital social and health care services?

Materials and Methods

Sample

The study is part of the large project examining digital exclusion of socially marginalized groups, including people with vulnerabilities in mental health who formed a purposive sample in this study. Participants ($N = 121$) for a cross-sectional survey were recruited between November 2020 and May 2020 through four organizations providing community-based mental health rehabilitation, support, and sheltered employment services for adults with mental health difficulties in southern Finland. The majority of participants (64%) were recruited from a medium-sized public services provider, and the sample was complemented from three non-governmental organizations (NGOs) providing rehabilitation and peer support to their members. The calculated response rate for the enterprise was approximately 25%. Participants from other organizations were individually invited to participate, and therefore their response rates remained unknown. Data collection was carried out with a questionnaire, and the participant had a choice to respond either in paper or digital form. Responses in paper form were returned anonymously in the provided envelopes.

The study was conducted according to the guidelines of the Finnish National Board on Research Integrity (2019), based on voluntary participation and informed consent. Ethical approval was obtained from the ethics committees for humanities at universities of Applied sciences in the Helsinki metropolitan area (Decision 6/2020; September 25, 2020). Mental health service users may experience cognitive problems; therefore, special effort was put into designing the questionnaire with a clear layout and simple language, and it was tested by several participants. NGO staff members were instructed to give advice on the questionnaire completion.

Measures

The questionnaire comprised several measures with 24 questions, most of which contained several items. Digital social and healthcare services referred to in this study are services such as online social welfare applications, remote appointments with professionals, use of the national patient data repository, use of health information portals, symptom assessments, peer support, and self-help programs offered mainly by governmental agencies, municipal authorities, and NGOs. In the questionnaire, examples of the most common digital social and health care services were provided to respondents.

This study focuses on 10 questions with 62 items. In addition, demographic questions including age, gender, level of education, and relationship status were explored. The respondents were asked about the use of social and health care services, use of the Internet, and the types of access methods and services they have at their disposal. Furthermore, respondents provided information on the support sources they used while accessing the Internet and digital services.

General self-efficacy was measured by 10 items on the Schwarzer and Jerusalem scale.⁴³ In the scale, the possible responses are: (1) not at all true, (2) hardly true, (3) moderately true, and (4) exactly true. The Internet self-efficacy scale was adopted from the Eastin and LaRose¹⁸ study, with eight items measuring the use of Internet hardware and software at a general level. Proxy efficacy was operationalized through eight items on professional, family, friend, and volunteer efficacy, with the efficacy referring to respondents' confidence in the ability of the aforementioned persons to provide support in using digital services.^{14,15,29,37,44} The 5-point Likert scale, ranging from 1 ("totally disagree") to 5 ("totally agree"), was applied to both proxy efficacy and Internet self-efficacy response options. The behavioral intention to use digital social and health care services was used to indicate the use behavior. Following previous studies in health technology^{21,24,27} and the theory of planned behavior,⁴⁵ a five-item measure for behavioral intention was planned. The first two items related to a person's overall readiness to use digital services and the likelihood of future use of digital services. Eagerness to apply for social welfare benefits or deal with health-related matters over the Internet was measured using two questions. The last item concerned preference between remote and face-to-face appointments with social and healthcare professionals.

Statistical Analysis

Descriptive statistics and analyses were computed using IBM SPSS software (version 27, IBM Corp.). Descriptive statistics were conducted to describe demographic characteristics and background variables of the study sample, Internet access, the use of digital products and services, and the support received from various actors (see [Tables 1](#) and [2](#)). Average indices were calculated for the measures of self-efficacy, Internet self-efficacy, proxy efficacy, and intention to use digital health and social care services, and their internal consistency was estimated with Cronbach's alpha (see [Table 3](#)).

Pearson's product-moment correlation coefficient was applied to examine the correlations between the main variables and age (see [Table 4](#)). The background variables of gender, relationship status, and educational level were dichotomized.

Table 1 Demographic Characteristics, Internet Access, Services, and Usage Among the Study Participants

Variable	Value	
Age (years), mean (SD), n = 107	43.9	(12.6)
Gender, n (%)		
Male	45	(37.5)
Female	76	(62.5)
Relationship status, n (%)		
Married/in a relationship	32	(26.4)
Single/unmarried	89	(73.6)
Educational level, n (%)		
Secondary education	86	(75.4)
No secondary education	28	(24.6)
Internet usage, n (%)		
Every day or almost every day	109	(90.1)
Weekly, monthly, or several times a year	9	(7.4)
Once a year or never	3	(2.5)
Places to access Internet, n (%)		
Home	114	(93.4)
Family, friends, and neighbors	71	(58.2)
Public places	63	(51.6)
NGOs, parish	32	(26.2)
Social and health care services and institutions	31	(25.4)
No access	1	(0.8)
Mobile and fixed Internet products, n (%)		
Smartphone	106	(86.9)
Personal computer	90	(73.8)
Cable or NBN fixed broadband connection	76	(62.3)
Tablet computer	41	(33.6)
Shared family computer	17	(13.9)
No Internet product	3	(2.5)
e-Identification methods, n (%)		
Online bank IDs	108	(88.5)
Mobile IDs	48	(39.3)
None	1	(0.8)
Health and social services used within past year, n (%)		
Social Insurance Institution services	105	(86.1)
Health care services	96	(78.7)
Mental health services	60	(49.2)
Municipal social welfare services	49	(40.2)
Substance abuse services	9	(7.4)
Family and social services	7	(5.7)

Table 2 Assistance Received for Using Digital Services

Variable	n	(%)
Family members	70	57.9
Friends, acquaintances, or relatives	57	47.1
No one — I solved the problem myself.	45	37.2
Social and health care professional	20	16.5
Digital support point in library	11	9.1
Organization (NGO)/ church employee or volunteer	11	9.1
I did not get any help, and I was not able to use the service.	11	9.1
City or municipal service point	8	6.6

Table 3 Construct Variables and Their Means, SDs, and Reliabilities

Variable	Items	N	Mean	SD	Cronbach α
Intention to use digital services	5	119	3.20	1.03	0.83
Proxy efficacy	8	118	3.61	0.76	0.77
Self-efficacy	10	114	2.79	0.61	0.92
Internet self-efficacy	8	116	3.67	1.04	0.94

Table 4 Correlations Among Main Variables

Variable		1	2	3
1. Intention to use digital services	<i>r</i>	— ^a	—	—
	<i>p</i> value	—	—	—
	N	—	—	—
2. Self-efficacy	<i>r</i>	0.05	—	—
	<i>p</i> value	0.58	—	—
	N	113	—	—
3. Internet self-efficacy	<i>r</i>	0.54	0.28	—
	<i>p</i> value	<0.001	0.003	—
	N	115	111	—
4. Proxy efficacy	<i>r</i>	0.49	0.09	0.28
	<i>p</i> value	<0.001	0.33	0.003
	N	117	114	114

Note: ^aNot applicable.

Subsequently, correlations between main and background variables were explored using the point-biserial correlation method. Linear regression models with the Enter method were then used to see whether self-efficacy, Internet self-efficacy, and proxy efficacy with background variables predict the intention to use digital social and health care services (see Table 5). Analyses were based on observations with no missing data (complete analyses). The validity of conditions was checked graphically with normality of residual distribution and linearity. Multicollinearity between the independent variables was explored using the VIF coefficients.

Moderation effects (age*self-efficacy, age*Internet self-efficacy) were preliminarily studied by regression analysis, but no significant effects were found. Mediation effects of Internet self-efficacy were further analyzed using Hayes' bootstrapping method. PROCESS macro 3.5.3 for SPSS was used for the mediation analysis, and 5000 bootstrap samples were used. The indirect effect was considered significant when the upper and lower bounds (95% confidence intervals) did not contain a value of zero.

Table 5 Linear Regression Analysis (Dependent Variable: Intention to Use Digital Services)

Model 1 ^a						
Independent Variable	β	SE	Beta	t	p value	VIF
(constant)	−0.29	0.76		−0.38	0.71	
Self-efficacy	−0.14	0.14	−0.09	−0.96	0.34	1.09
Internet self-efficacy	0.47	0.12	0.43	3.94	<0.001	1.63
Proxy efficacy	0.45	0.12	0.35	3.77	<0.001	1.18
Age	0.01	0.01	0.07	0.67	0.50	1.38
Gender [†]	0.17	0.18	0.08	0.96	0.34	1.07
Relationship status [†]	0.12	0.20	0.05	0.60	0.55	1.05
Education [†]	0.23	0.21	0.09	1.07	0.29	1.07
Model 2 ^b						
Independent Variable	β	SE	Beta	t	p value	VIF
(constant)	0.04	0.40		0.10	0.92	
Internet self-efficacy	0.40	0.08	0.40	5.02	<0.001	1.10
Proxy efficacy	0.48	0.10	0.37	4.72	<0.001	1.10

Notes: ^aR² = 0.385, Adjusted R² = 0.334, F = 7.6; p < 0.001; N = 93.

^bR² = 0.382, Adjusted R² = 0.371, F = 34.1; p < 0.001; N = 113. [†]Dummy.

Results

Respondents and Their Background

Respondents were welcome to choose more than one option when responding to questions regarding their Internet use, access, services, and products, and recent use of digital social and health care services. Participants (N = 121) had a mean age of 43.9 (SD = 10.4) years, over half were female (62.5%), and most of them (75.4%) had pursued secondary education. The vast majority had used social insurance services (86.1%) and health care services (78.7%) within the past year. Over 90% of participants used the Internet every day or almost every day, and for Internet access, 86.9% had a smartphone at their disposal. (See Table 1 for additional details on participants' background.)

The main source for support in using digital social and health care services were family (57.9%) and friends (47.1%) (see Table 2). Of the respondents, 16.5% had received help from social and health care professionals, 9.1% from a library or NGO, and 6.6% from municipal service points. However, on many occasions, respondents had solved the problems by themselves (37.2%), and 9.1% reported that they did not receive help from anyone, resulting in a failure to use the service.

Scale Internal Consistencies

The internal consistencies of intention to use digital services, general self-efficacy, proxy efficacy, and Internet self-efficacy scales were assessed by calculating Cronbach's alpha (Table 3). This analysis indicated that the intention to use digital services, general self-efficacy, and Internet self-efficacy measures had adequate internal consistency with alpha scores of 0.83, 0.92, and 0.94, respectively. The proxy efficacy demonstrated acceptable properties with an internal consistency of Cronbach's $\alpha = 0.77$.

Intention to Use Digital Social and Health Care Services

When participants were asked if they would use digital social and health services in the future, the vast majority (68.8%) of respondents partially or totally agree. Primarily accessing social and health care services through digital channels was preferred by slightly over half of the respondents (56.5%). When needing to talk to social and health care professionals, only 22.9% of the respondents would partially or totally agree on choosing a remote appointment over a face-to-face meeting (Table 6).

Table 6 The Distributions of Responses Regarding Behavioral Intentions

Statements	N	I Totally Disagree (%)	I Partially Disagree (%)	I Neither Agree nor Disagree (%)	I Partially Agree (%)	I Totally Agree (%)
I will use digital social and health services whenever possible in the future.	121	8.2	9	13.1	33.6	35.2
I am likely to primarily deal with social and health services in electronic form in the future.	120	11.5	9	21.3	30.3	26.2
I want to manage matters related to my social welfare benefits via the Internet in the future.	121	13.1	5.7	17.2	27	36.1
I would like to primarily take care of my health-related issues via the Internet in the future.	121	27	15.6	22.1	23.8	10.7
When I need to talk to a professional (social worker, doctor, nurse, etc.) in the future, if possible, I will prefer to meet remotely rather than face-to-face.	120	41.8	18	15.6	13.1	9.8

Correlations

The correlations between main variables and age were examined using Pearson's correlation coefficient. The intention to use digital social and health care services correlated significantly with proxy efficacy, $r(115) = 0.49$; $p < 0.001$, and Internet self-efficacy, $r(117) = 0.49$; $p < 0.001$. Self-efficacy did not correlate significantly with the intention to use digital services. However, self-efficacy and Internet self-efficacy correlated significantly with each other $r(111) = 0.28$; $p = 0.003$. Significant correlation was also found between proxy efficacy and Internet self-efficacy, $r(114) = 0.28$; $p = 0.003$ (Table 4). Age was negatively correlated with the intention to use services and Internet self-efficacy. Younger adults were more confident in their Internet skills, $r(103) = -0.50$; $p < 0.001$, and more inclined to use digital services, $r(106) = -0.25$; $p = 0.01$. Compared to women, men feel more confident in their Internet skills, $r(116) = -0.25$; $p = 0.007$. Furthermore, being in a relationship was positively related to Internet skills, $r(116) = 0.19$; $p = 0.04$. A higher educational level was positively related to Internet self-efficacy, $r(109) = 0.28$; $p = 0.004$ and the intention to use digital services, $r(112) = 0.23$; $p = 0.01$.

Linear Regression Analysis

Factors that explain the intention to use digital social and health care services were examined with linear regression analysis. Independent variables in the analysis were self-efficacy, proxy efficacy, and Internet self-efficacy. Relevant background variables such as age, gender, relationship status, and level of education were included in the analysis. The VIF coefficients ranged from 1.05 to 1.63, showing no excessive multicollinearity in the regression analysis. The first regression model (see Table 5) includes all the independent variables examined. None of the demographics were significant and were therefore not included in the second model. Both models explain perceived control quite well ($R^2 = 0.385/0.382$). The analyses show that both Internet self-efficacy and proxy efficacy explain the intention to use digital social and health care services.

According to regression analysis (Table 5), the effect of self-efficacy was not significant, and self-efficacy did not moderate the effect of other independent variables. Mediation analysis using the bootstrap method revealed that self-efficacy has an indirect positive effect on the intention to use digital services, and Internet self-efficacy was the mediating variable (coefficient 0.22; 95% CI 0.08–0.39; $p = 0.002$).

Discussion

The first aim of this cross-sectional study was to explore the relationship between proxy efficacy, general self-efficacy, Internet self-efficacy, and the use of digital social and healthcare services among adult mental health service users within four Finnish NGOs. The second aim was to find out sources of social support for using digital services.

To understand the characteristics of the purposive sample, several background variables were examined. Of the respondents, 76 were women and 45 were men, the average age was 43.9 years, and 75.4% had secondary education. Over 90% of the participants used the Internet every day or almost every day, mostly at home. The majority of respondents had a smartphone and personal computer at their disposal and possessed online bank IDs required for using digital services in Finland. To summarize, respondents were well-equipped with ICT, reflecting the current situation in Finland, where 91% of adults use a smartphone and 80% a personal computer daily.⁴⁶ Respondents also reported relatively high proxy efficacy and Internet self-efficacy beliefs, and accordingly, over one-third of them reported that they had been able to solve problems encountered using digital services by themselves. A cautious conclusion can be drawn that a lack of digital devices or general ICT skills does not pose a serious risk of digital exclusion among respondents despite their vulnerable backgrounds, cf.³ Thus, the contributions of this study lie in adding understanding to the role of self-efficacy, Internet self-efficacy, and proxy efficacy in explaining the use of digital services.

According to the findings, respondents' willingness to use digital services was significantly affected by their Internet efficacy and proxy efficacy beliefs. General self-efficacy beliefs were not directly related to the intention to use the services, supporting previous claims that self-efficacy should be assessed by the targeted performance,²² and domain-specific self-efficacy holds better predictive power than general self-efficacy.²³ However, previous studies^{25,26} suggest that self-efficacy has an indirect positive effect on intention. The mediation analysis results confirmed that self-efficacy beliefs enhanced Internet self-efficacy. Overall, the present findings are consistent with the previous findings that Internet self-efficacy as task-specific efficacy enhances the use of health-related digital services.^{21,24} The assumption^{12,29} that proxy efficacy beliefs and self-efficacy beliefs reinforce each other was not supported. These findings suggest that to enhance service users' confidence in using digital services, reliance on Internet self-efficacy and proxy efficacy, rather than general self-efficacy beliefs, should be encouraged.

In accordance with previous studies, proxy efficacy beliefs¹⁷ and social support^{16,30} predicted the intention to use digital services. The respondents received assistance mostly from close family members, relatives, and friends, whereas assistance from official actors remained at a low level. Strikingly, social and health care professionals provided support to only 16.5% of respondents, while the recent Act on the Provision of Digital Services (2019)⁴⁷ obligates public authorities to provide support for service users. Furthermore, the findings show that the third sector (ie, volunteers, NGOs, and parishes) had only a supplementary role in digital support for mental health service users as a vulnerable group, in contrast with the assumptions within the operating model for digital support in Finland.⁴⁸ Instead, the findings emphasize the role of "warm experts", ie, family members and other close ones, in giving support, consistent with previous studies.^{16,30,34,40,49}

According to Bakardjieva,¹⁶ warm experts are unique for their in-depth knowledge of the needs of novice users and are thus able to give appropriate instructional and emotional support to them. For example, when older adults have potential support close at hand, such as a partner who uses technology, they are more likely to be Internet users.⁵⁰ The quality of proxy agents is an important factor for engaging individuals with new behaviors.^{14,15} On the other hand, support by warm experts can compensate for deficiencies in individuals' competence and self-efficacy; but on the other, there may also be a risk of over-reliance on others with regard to digital matters.^{14,38,39} Proxy agents need to find ways to foster proxy efficacy while simultaneously developing self-reliance.

Overall, participants in this study were willing to use digital social and health care services in the future. Two-thirds of the respondents would use digital services whenever possible and over half would primarily take care of their social and health care affairs via the Internet. This is consistent with previous findings that mental health service users benefit from digital services, and they are generally inclined to use them.^{3,4,51} However, regarding the mode of service, most respondents (59.8%) did prefer face-to-face appointments when they need to talk to professionals. Similarly, vulnerable young people with more advanced digital skills prefer face-to-face discussions with professionals when talking about difficult personal issues.⁵² When dealing with complex issues, digital welfare services seem to supplement rather than substitute face-to-face appointments.^{10,53} A preference for face-to-face interactions is also documented among prisoners.⁵⁴ Moreover, mental health illnesses may induce symptoms that affect cognitive skills and make it difficult to achieve a therapeutic relationship with professionals via remote appointments.^{3,4} People with mental health problems

may have concerns about using web-based services and would rather choose appointments with the professionals in person.⁵⁵

Taken together, these findings suggest that while online contacts might be appropriate for running regular businesses, online contacts cannot replace face-to-face contacts when issues are difficult and complicated, such as welfare applications with multiple required attachments or which involve deeply personal matters, such as life crises.

The digital transformation is a constant state where technological advances are rapidly emerging. Therefore, vulnerable groups need continuous support to prevent their digital exclusion. Proxy agents may increase motivation to use the web-based services. This study points out that close people are key supporters, and therefore, more attention should be paid to the quality of proxy support and its role alongside other forms of support.^{49,56} Further research on efficacy beliefs is crucial. Exploring whether informal support sources benefit self-efficacy, and whether informal and formal support differs in terms of the effect on self-efficacy, may offer guidance in determining what kind of support could be most useful. The quality dimensions of proxy agency, especially within the mental health context, would also benefit from closer scrutiny as support needs vary.

Limitations

The study has been carried out in four organizations providing community-based mental health services that mainly operate in southern Finland. The COVID-19 pandemic substantially hindered data collection when researchers' face-to-face contacts with service users were restricted, thus making it harder to reach target groups and subsequently affecting the sample diversity. It is plausible that service users with digital devices and more advanced digital skills were more motivated to participate in the study when instructed by staff members. However, the questionnaire was offered in paper and electronic format, which provided the opportunity to participate even without Internet access. It is also worth pointing out that only 49% of respondents reported having used mental health services within the past year, even though at the same time, they were users of various community-based mental health services, indicating that participants' mental health problems were also varied. The sample represented individuals who participated in rehabilitative services within a community setting and various peer group activities. Overall, the group of respondents seemed to possess the adequate functional capacity to use digital devices and services. Hence, it is reasonable to assume that the survey did not reach the individuals with the most debilitating constellation of symptoms that potentially prevents the use of digital services. The survey, which was in Finnish, also excluded potential non-Finnish speaking participants, even though some of them did fill in the questionnaire with staff members' support. Some degree of bias is also evident because the most underprivileged individuals in need of services are beyond reach, meaning they are unable to provide valuable information on their challenges in a digital environment.

Together with the small sample size, the aforementioned biases restrict the generalizability of the results to a wider number of mental health service users. Moreover, it was found that Internet self-efficacy and proxy efficacy predicted the intention to use digital health care and social welfare services, but due to the cross-sectional data, causal relationships cannot be confirmed. In the future, the effects of proxy efficacy on digital skills need to be examined using a longitudinal design.

Conclusion

Mental health service users constitute a group of people who are at greater risk of digital exclusion. Internet self-efficacy and proxy efficacy are major contributors to the intention to use digital social and health care services. The need for support from family, friends, volunteers, and professionals is particularly obvious. The support received from "warm experts" (ie, family and friends) was significantly higher than the support from other sources. Proxy agents may serve as facilitators to reduce the digital divide and promote the use of digital services by complementing individuals' self-efficacy and Internet self-efficacy beliefs. However, proxy agents should be aware of the over-reliance factor and foster individuals' self-reliance throughout the process. Peer and formal support should be addressed to enhance the quality of proxy efficacy. It is important to focus on the supply and quality of sources of support to ensure adequate coverage and tailored support for individual needs. Overall, proxy agents would benefit from structural education on guidance practices to develop digital support towards a uniform level of quality.

Abbreviations

ICT, Information and communications technology; NGO, Non-governmental organization.

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Disclosure

Dr Soile Juujärvi reports personal fees from Edistia Group outside the submitted work. She is also a member of the advisory board of one of the organizations involved in the study. However, the study has not been on its agenda. The authors report no other conflicts of interest in this work.

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