
Review

Efficacy of Technology-Based Interventions on the Reduction of Loneliness: Systematic Review and Meta-Analysis

Zdenek Meier, PhD; Marie Buchtova, PhD; Jan Sandora, MA; Lukas Novak, PhD; Jakub Helvich, PhD; Ondrej Buchta, MA; Jana Furstova, PhD; Klara Malinakova, PhD; Peter Tavel, DProf

Olomouc University Social Health Institute, Palacký University Olomouc, Olomouc, Czech Republic

Corresponding Author:

Zdenek Meier, PhD
Olomouc University Social Health Institute
Palacký University Olomouc
Univerzitni 244/22
Olomouc 771 11
Czech Republic
Phone: 420 585637394
Email: zdenek.meier@oushi.upol.cz

Abstract

Background: Loneliness is a widespread public health concern linked to increased risks of health problems. As populations age, the demand for effective interventions to mitigate loneliness continues to grow.

Objective: This meta-analysis aimed to examine the effectiveness of technology-based interventions in reducing loneliness.

Methods: A systematic literature search was conducted in Web of Science, PsycInfo, PubMed, Scopus, Google Scholar, Embase, and the Cochrane Library in August 2024. We included randomized controlled trials that examined the effectiveness of technology-based interventions compared with any control group in reducing loneliness across all age groups. Nonrandomized studies, qualitative research, and studies lacking sufficient statistical data for effect size calculation were excluded. After screening 1089 records, 7 studies involving 580 participants met the inclusion criteria. Data were extracted by 3 independent reviewers, with discrepancies resolved by a fourth reviewer. The risk of bias was assessed using the Cochrane risk-of-bias tool. A random-effects model was used to synthesize effect estimates, with standardized mean differences as the primary effect size metric. Heterogeneity was assessed using the Q statistic and I^2 index, and a prediction interval was calculated to estimate the expected range of true effects.

Results: We found a small and statistically nonsignificant overall effect of technology-based interventions on loneliness (pooled standardized mean difference = -0.21, 95% CI -0.59 to 0.17; 95% prediction interval -1.14 to 0.63). Substantial between-study variability was present ($I^2=57%$; $\tau^2=0.08$; $\tau=0.28$), and the prediction interval indicated that true effects in future studies may range from substantial reductions to moderate increases in loneliness. Differences between intervention types could not be examined due to the limited number of eligible studies. The Egger test showed significant funnel plot asymmetry ($t_5=-5.47$; $P=.003$). However, since the trim-and-fill method identified no missing studies, the asymmetry is unlikely to be fully explained by publication bias.

Conclusions: Unlike previous reviews, which focused primarily on older populations or included nonrandomized studies, this meta-analysis provides a rigorous synthesis of only randomized controlled trials across the lifespan. Our findings do not provide evidence for a reliable reduction in loneliness following technology-based interventions. Moderate heterogeneity indicates that effects differ across studies, and the limited number of eligible trials prevented the analysis of potential moderators. By incorporating robust prediction intervals and a broader age demographic, this study offers a more comprehensive view of the variability of intervention outcomes compared with existing literature. The wide prediction interval suggests that intervention effects may vary across settings. In practice, technology-based approaches should be viewed as supportive tools that require careful tailoring rather than universal solutions to loneliness.

Trial Registration: PROSPERO CRD42024505117; <https://www.crd.york.ac.uk/PROSPERO/view/CRD42024505117>

J Med Internet Res 2026;28:e80059; doi: [10.2196/80059](https://doi.org/10.2196/80059)

Keywords: loneliness; technology; intervention; meta-analysis; systematic review

Introduction

Negative Impacts of Loneliness

Loneliness is a growing public health challenge linked to a wide range of adverse mental and physical health outcomes. It has gained considerable attention in recent years due to its significant impact on both mental and somatic health. It is a major risk factor for various physical health problems, influencing biological responses to stress [1,2], glucose regulation [3], cardiovascular risk [4,5], and overall health behaviors [6]. Furthermore, loneliness is a critical risk factor for mental health, as individuals experiencing loneliness exhibit higher levels of psychological distress, depression, and poorer subjective health assessments [7]. It is associated with an increased prevalence of paranoid symptoms [8], depression [9,10], anxiety [10,11], and psychotic symptoms [8], as well as lower self-esteem and self-evaluation [12,13]. A study by Liang et al [14] identified loneliness as a risk factor for more than two-thirds of the analyzed diseases and for 13 out of 14 disease categories, particularly mental and behavioral disorders, infectious diseases, and conditions affecting the nervous, respiratory, and endocrine systems [14]. Lastly, loneliness is associated with an increased risk of mortality [15,16].

Loneliness encompasses both adaptive and maladaptive aspects. In the cognitive model, short-term loneliness can function as a social mechanism, prompting individuals to strengthen interpersonal bonds and seek social connections [17]. This adaptive process can be considered an evolutionary response to the fundamental human need for belonging [18]. However, chronic loneliness leads to negative consequences affecting both physical and mental health [17,19,20]. The model of loneliness as an adaptive mechanism with maladaptive consequences is essential not only for understanding its biological function but also for recognizing the risks associated with its long-term presence in an individual's life. Loneliness can be helpful in the short term, but chronic loneliness leads to poorer health and reduced life satisfaction [21].

Loneliness has widespread societal impacts, including economic consequences such as increased health care costs and reduced work productivity [22]. Effective solutions to loneliness lie in future research and interventions aimed at preventing its occurrence and mitigating its negative effects on both individuals and society [23-25]. As a significant public health issue, loneliness requires a multidisciplinary approach involving collaboration among experts in health care, psychology, and social sciences [24,26,27]. Key strategies for addressing loneliness are particularly found in social work and social policies, which can help reduce risk factors and promote social inclusion [26,28].

Definition

From a multidimensional perspective, loneliness can be defined as a significant psychological and social

phenomenon, primarily conceptualized in the literature as the subjective perception of a discrepancy between expected and actual social relationships [29]. This discrepancy is not necessarily objectively measurable by the number of social interactions but rather by the qualitative evaluation of one's relationships [30]. Moreover, loneliness is a complex phenomenon encompassing both affective and cognitive components. The affective component relates to the emotional distress associated with feelings of isolation, whereas the cognitive component reflects the perceived inadequacy of social relationships [21].

Weiss [31] distinguished between social loneliness, which is related to a lack of social networks and a sense of social exclusion, and emotional loneliness, which is associated with the absence of close and meaningful relationships [31]. In addition to these, another critical dimension is existential loneliness, characterized by a profound sense of isolation linked to questions of life's meaning, human finiteness, and the subjective incommunicability of individual experience [32].

Based on these observations, loneliness can be explored according to its 3 dimensions: intimate loneliness, which pertains to the lack of a deep relationship with a close person; relational loneliness, which arises from the absence of quality friendships or family connections; and collective loneliness, which stems from a missing sense of belonging to a group or a broader community [33-35]. The concept of collective loneliness remains a subject of debate and requires further investigation and clarification [36].

When examining the relationship between loneliness and the quality and quantity of social relationships, particularly among older adults, it is essential to consider not only quantitative aspects, such as the number of friends and frequency of interactions, but also the perception of social support and emotional closeness, which determine the quality of relationships [30].

The Prevalence of Loneliness

Loneliness is highly prevalent across the lifespan and in diverse sociodemographic groups [7-9]. Population-based studies and meta-analyses indicate that roughly 1 quarter to one third of adolescents and older adults report at least moderate levels of loneliness, with substantial variation between countries and regions [37-48]. Although some longitudinal studies suggest a U-shaped or more complex trajectory, with elevated loneliness particularly during adolescence and in later life [37,49-51], these patterns appear to be driven by life transitions and changes in social roles rather than age per se [40,41,49]. Rates of loneliness increased further during the COVID-19 pandemic, when large-scale public health measures such as social distancing and lockdowns disrupted everyday social contact [52-56].

Chronic loneliness is especially concerning; estimates suggest that about 1 in 5 older adults experience persistent loneliness, which is most strongly associated with adverse

physical and mental health outcomes [47,57]. Against this backdrop, there is an urgent need for scalable, evidence-based interventions that can be implemented across different age groups and settings.

Technology-Based Interventions for Loneliness

In recent years, digital technology has emerged as a promising avenue for delivering loneliness interventions. Technology-based interventions are approaches that use digital or communication technologies as the primary mode of delivery—for example, web-based programs, mobile apps, videoconferencing platforms, or socially assistive robots designed to enhance social connection or provide psychosocial support [58]. These interventions can extend the reach of traditional face-to-face programs, particularly for individuals who are homebound, geographically isolated, or affected by situations such as the COVID-19 pandemic, where physical distancing measures increased both loneliness and the need for remote support [58,59]. By enabling people to connect with peers, family members, or professional services via digital means, they offer a potentially scalable way to address loneliness [59,60]. However, the existing evidence base remains limited and heterogeneous, with many small trials of variable quality and relatively few rigorous randomized controlled studies [58].

Existing systematic reviews of loneliness interventions have often focused on traditional face-to-face programs or on specific subgroups such as older adults, and several recent reviews of digital interventions report small or nonsignificant effects, substantial heterogeneity, and methodological limitations. As new trials using diverse technologies and age groups have accumulated, an updated and methodologically rigorous synthesis is needed to clarify under what conditions technology-based interventions can meaningfully reduce loneliness.

Thus, this study aims to conduct a systematic review and meta-analysis to evaluate the effectiveness of technology-based interventions compared with control conditions in reducing loneliness among individuals across the lifespan. We focus on interventions in which digital or communication technologies constitute the main delivery channel (eg, mobile apps, web-based self-help programs, videoconferencing, socially assistive robots). Specifically, we address the following research questions:

1. To what extent do technology-based interventions reduce loneliness compared with control conditions?
2. Do the effect sizes differ between distinct types of technology-based interventions (eg, app-based psychological programs, socially interactive technologies, videoconferencing-based approaches)?
3. Are intervention or sample characteristics such as participant age, intervention duration, and follow-up period associated with differences in effectiveness?

Clarifying these questions is particularly important in the context of population aging and growing shortages in health and social care workforces, where scalable digital solutions

may help to complement, but not replace, traditional forms of support.

Methods

Research Design

The protocol for this study was registered with the PROSPERO (International Prospective Register of Systematic Reviews) under the registration number CRD42024505117 [61]. The systematic review and meta-analysis were conducted and reported in accordance with the PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analyses) statement [62,63], using the PRISMA 2020 expanded checklist to ensure comprehensive reporting. The literature search was carried out following the PRISMA-S extension for reporting search strategies [64], and the abstract was structured according to the PRISMA 2020 for Abstracts guidelines. Fully completed versions of all relevant PRISMA checklists are provided in [Checklist 1](#).

Eligibility Criteria

To ensure methodological rigor, this meta-analysis included studies meeting the following criteria: (1) no restrictions on publication language; (2) no restrictions regarding the age of participants; (3) empirical studies using a randomized controlled trial (RCT) design; (4) use of a technology-based intervention compared with a control group; and (5) reporting of loneliness as a measured outcome (either as a primary or secondary end point), with available loneliness levels before and after the intervention or sufficient data to calculate the effect size. Furthermore, no restrictions were placed on the study setting (eg, community-dwelling or long-term care facilities) or the minimum duration of the intervention. Regarding the definition of loneliness, studies were eligible if they operationalized the construct consistent with the conceptual framework described in the *Introduction* section, assessing subjective feelings of loneliness or social isolation using quantitative measures. Studies that did not meet all of these criteria, including nonempirical studies, non-RCT designs, interventions unrelated to technology, and those lacking relevant outcome measures, were excluded.

Search Strategy

The literature search for this meta-analysis was conducted in October 2023, with a final update performed on August 21, 2024. The following databases were systematically searched from their respective inception dates: Web of Science (Clarivate), PsycInfo (EBSCOhost), PubMed (NCBI), Scopus (Elsevier), Google Scholar, Embase (Elsevier), and the Cochrane Library (Wiley). To identify relevant studies, search terms were organized into four thematic groups: (1) loneliness (eg, “loneliness,” “perceived social isolation”); (2) intervention (eg, “intervention,” “treatment”); (3) technology (eg, “technology,” “computer,” “mobile,” “app*”); and (4) study design (eg, “Randomized Controlled Trial,” “RCT”). Searches were conducted in the fields of Title, Abstract, and Keywords, except in Google Scholar, which did not support specific operators for limiting searches to these

fields. Boolean operators and truncation symbols were used to ensure a comprehensive and systematic identification of studies relevant to technology-based interventions for reducing loneliness. The exact search syntaxes applied to each database are detailed in [Multimedia Appendix 1](#). In addition to database searches, further studies were identified through citation searching in the reference lists of relevant review articles on the topic (see details below).

Selection Process

All retrieved citations were imported into Zotero reference management software [65]. Duplicate records were identified and removed using the software's automated functions, followed by manual verification. Titles and abstracts underwent an initial screening to determine relevance. First, records that did not report published results of primary empirical research were excluded. This included books, study protocols, preprints, review articles (including meta-analyses), theoretical studies, conference contributions, master's theses, commentaries, editorials, letters to the editor, and studies with inaccessible full texts. Initially, we also intended to exclude dissertation theses. However, we decided to retain them in the search, as we consider the dissertation review process to be sufficiently rigorous. Second, studies that were thematically irrelevant were excluded. This applied to records that did not focus on loneliness (eg, loneliness was not measured as an outcome) or did not examine the impact of technology-based approaches (eg, the intervention was not technology-driven, or the study did not compare nontechnology vs technology-based methods). Third, studies that did not follow an RCT design were excluded. This included, for example, studies assessing the feasibility of RCTs, those using qualitative designs, or other nonexperimental approaches.

Thematically relevant reviews were used for further citation searching to help identify additional empirical studies not captured in the initial database search. The full texts of studies were reviewed based on predefined inclusion and exclusion criteria. The selection process was conducted by 3 reviewers. Each record and report was screened independently by at least 2 reviewers. Any discrepancies were resolved through consultation with a fourth reviewer. All stages of the study selection process were conducted manually; no automation tools or machine learning classifiers were used.

There were minor deviations from the original PROSPERO protocol. First, the literature search was updated in August 2024 to include the most recent evidence, extending the search period beyond the date initially planned. Second, while the protocol mentioned a dedicated search for gray literature, the final strategy relied on comprehensive database searches and citation tracking of relevant reviews. Finally, the GRADE (Grading of Recommendations Assessment, Development, and Evaluation) approach and the calculation of prediction intervals were added during the analysis stage to provide a more robust assessment of evidence certainty and effect distribution.

Data Collection Process

A predefined data extraction form was used to collect the following information: (author, year, country), participant demographics (sample size, characteristics, age range, mean age, and gender distribution), baseline loneliness levels, measurement instruments, and intervention details (type of intervention and control, duration, and the timing of postintervention assessment). For all included studies, loneliness data were extracted using the specific quantitative measure provided by the authors. In cases where studies reported results from multiple postintervention time points, data from the assessment conducted immediately following the completion of the intervention were extracted to ensure consistency. Data extraction was conducted by 3 reviewers, with each study independently coded by 2 reviewers. Any discrepancies were resolved by involving a fourth reviewer. The process was performed manually, and no automation tools or machine learning applications were used. All relevant data were obtained through a comprehensive review of the text, tables, and figures of each report; no data extraction software for figures was required. As all included studies were published in English, no formal translation process was necessary. Only studies reporting sufficient statistical information to compute pre-post effect sizes were included. Thus, according to this criterion, 8 studies were excluded. We chose not to contact study authors for missing data because of practical constraints and the likelihood of delayed or incomplete responses. To ensure transparency and replicability, we decided that only studies with fully reported data necessary for effect size computation (including pre-post correlations) were included. This approach allows the meta-analysis to rely only on verifiable and publicly available information.

Data Synthesis

A random-effects model was used to synthesize effect estimates across studies, accounting for between-study heterogeneity. Effect sizes and their sampling variances were calculated as standardized mean differences (SMDs), specifically the difference between standardized mean changes in the intervention and control groups. This metric was chosen to accurately account for the pre-post study design and to enable the comparison of results from studies using different loneliness measurement scales [66]. The magnitude of the pooled effect size was interpreted according to Cohen [67] conventional criteria: 0.2 (small), 0.5 (medium), and 0.8 (large). All computations were performed in R using the metafor package. The summary effect size and its 95% CI were computed to estimate the overall intervention effect. A random-effects model was fitted using restricted maximum likelihood, and the Hartung-Knapp-Sidik-Jonkman [68] adjustment was applied to obtain robust confidence intervals given the small number of studies ($k=7$). Additionally, a prediction interval was calculated using the bootstrap confidence distribution approach proposed by Nagashima et al [69] to define the expected range of true effects in comparable populations and future studies, as the prediction interval quantifies the range within which the true effect sizes

are expected to fall, providing a clearer interpretation of effect variation. Heterogeneity was assessed using the Q statistic and I^2 . The Q test was performed to assess the null hypothesis of homogeneity (ie, that all studies shared a common effect size), with a significance threshold of $\alpha=.10$ to account for its typically low statistical power. I^2 quantified the proportion of observed variance attributable to true heterogeneity rather than random error. However, as the I^2 index does not describe the magnitude of variation in effect sizes, we place primary emphasis on the prediction interval to interpret the distribution of true effects across different populations and settings. Given that estimates of heterogeneity based on fewer than 10 studies may be unreliable, the results should be interpreted with caution. Subgroup analysis and meta-regression were not conducted because the number of available studies was insufficient for reliable moderator testing. Egger regression test and Duval and Tweedie trim and fill procedure were conducted to examine potential dissemination bias. Potential publication bias was assessed within the same random-effects framework in R. To assess the robustness of the findings, a leave-one-out sensitivity analysis was conducted, in which the overall effect was recalculated after omitting each study in turn.

The risk of bias in the included studies was assessed using the revised Cochrane risk-of-bias 2 tool for randomized trials [70]. Three reviewers participated in the assessment process. Each study was evaluated across five domains: (1) bias arising from the randomization process, (2) bias due to deviations from intended interventions, (3) bias due to missing outcome data, (4) bias in measurement of the outcome, and (5) bias in selection of the reported result. Any discrepancies were resolved through joint discussion until a

consensus was reached. The overall risk-of-bias judgment for each study followed the standard Cochrane risk-of-bias 2 algorithm, where the highest risk level identified in any individual domain determined the overall rating. The visualization of the risk of bias assessment was produced using the *robvis* tool [71].

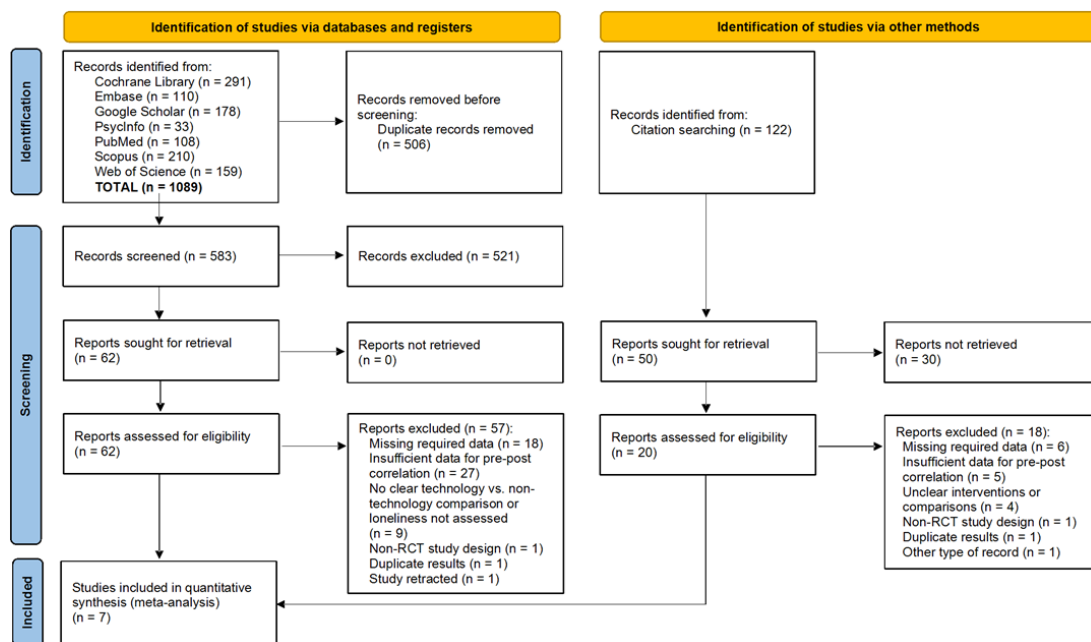
The certainty of evidence for the primary outcome (loneliness) was assessed using the GRADE approach [72]. Two reviewers independently performed the assessment, with any discrepancies resolved through discussion until a consensus was reached. We evaluated 5 domains: risk of bias, inconsistency, indirectness, imprecision, and publication bias. The certainty of evidence was graded as high, moderate, low, or very low, reflecting the degree of confidence that the true effect lies close to the estimate. The findings were summarized following the standard GRADE Summary of Findings table format.

Results

Study Selection

The database search yielded 1089 records. After screening titles and abstracts, full-text reports were assessed for eligibility. Detailed numbers of excluded records at each stage are presented in the PRISMA flow diagram (Figure 1). A total of 57 studies were excluded, primarily due to incomplete or insufficient data. A full list of excluded studies with reasons is provided in Multimedia Appendix 2. Notably, the study by Sun [73], which was initially considered, was excluded due to its retraction by the publisher [74].

Figure 1. PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analyses) flowchart illustrating the study selection process for the systematic review and meta-analysis of technology-based interventions for loneliness (literature search conducted up to August 2024). RCT: randomized controlled trial.



Additional studies were sought through citation searching of thematically relevant reviews (n=32). This search yielded 122 records, of which 72 duplicates were removed. Following the assessment of the remaining reports, most were excluded for reasons similar to the database search results, ultimately leaving 2 studies for inclusion (Figure 1). Moreover, to ensure the comprehensiveness of our study selection, we compared our included studies with those identified in the meta-analysis by Jin et al [75], which examined the effectiveness of technology-based interventions for reducing loneliness in older adults. Given the overlap in scope, we reviewed all 6 studies included in their meta-analysis. However, none of them met our strict inclusion criteria. Notably, although the study by Shapira et al [76] was classified as an RCT in the review by Jin et al [75], our assessment revealed that the allocation was based on participants' willingness and interest rather than randomization. Therefore, it was excluded as a non-RCT. Four other studies [77-80] were excluded due to lacking essential source data or statistical indicators needed for calculating pre-post correlations. Similarly, the study of Slegers et al [81] lacked a clearly defined comparison between a technology-based and a nontechnology-based intervention.

Importantly, the study by Schwindenhammer [82], identified through citation searching, was originally described

by the author as having a quasi-experimental design. However, a detailed examination of the methodology revealed that the assignment of nursing homes to intervention or control groups was determined by a random drawing. Consequently, this study qualifies as a cluster-RCT rather than a nonrandomized quasi-experiment. Since the study satisfied the randomization requirement and provided the necessary statistical indicators for calculating pre-post correlations, it was included in the final analysis. Ultimately, 7 studies were included in the quantitative synthesis.

Study Characteristics

Tables 1 and 2 summarize the characteristics of the 7 included studies, including participant demographics, intervention details, descriptive statistics, and methodological features. The studies included in the meta-analysis were published between 2006 and 2024 and were conducted in the United States (n=3), Switzerland (n=1), Poland (n=1), New Zealand (n=1), and Israel (n=1). The target populations were diverse, ranging from high school students [83] and young adults [84] to adults [85], chronically ill rural women [86], and older adults residing in the community or nursing homes [82,87,88]. Participant ages ranged from 14 to 100 years.

Table 1. Characteristics of included studies: participant demographics and settings.

Study	Country	Sample characteristics	Sample N (IG ^a /CG ^b)	Age range (y)	Age, mean (SD)	Gender (% female)
Brog et al [85]	Switzerland	Adults with at least mild depressive symptoms	Randomized: 107; analyzed: 95	18-81	40.36 (14.59)	81.3
Hill et al [86]	United States	Chronically ill rural women	Randomized: 120; analyzed: 100	35-65	N/A	100
Iyer et al [83]	United States	High school students	Randomized: 110; analyzed: 108	14-19	16.05 (1.34) ^c	81.5
Karkosz et al [84]	Poland	Young adults with subclinical depression or anxiety	Randomized: 81; analyzed: 81	18-35	25.7 (4.59) ^c	71.6
Robinson et al [87]	New Zealand	Older adults in hospital and rest home care	Randomized: 40; analyzed: 34	55-100	N/A ^d	67.5
Shapira et al [88]	Israel	Older community-dwelling adults	Randomized: 86; analyzed: 82	65-90	72 (5.63)	80.5
Schwindenhammer [82]	United States	Older adults in nursing homes	Cluster-randomized: 5 nursing homes (83 residents); analyzed: 80	71-97	86.12 (5.28)	87.5

^aIG: intervention group.

^bCG: control group.

^cComputed from the dataset available, not provided in the study itself.

^dNot applicable.

Table 2. Characteristics of included studies: intervention details and descriptive outcomes.

Study	Baseline		Postintervention		Measurement instrument	Intervention type	Control type	Intervention duration (wk)	Difference between baseline and follow-up measurement
	N	Mean (SD)	N	Mean (SD)					
Brog et al [85]	IG ^a : 53; CG ^b : 54	IG: 21.26 (4.82); CG: 20.37 (4.25)	IG: 43; CG: 52	IG: 19.88 (4.56); CG: 20.27 (4.04)	UCLA ^c (9-item)	Internet-based self-help program for psychological distress (CBT) ^d	Wait-list control (care as usual)	3	3 wk
Hill et al [86]	IG: 43; CG: 57	IG: 45.73 (9.99); CG: 43.97 (10.53)	IG: 43; CG: 57	IG: 43.15 (8.69); CG: 43.11 (10.94)	UCLA (20-item)	Online, asynchronous, peer-led support group and health teaching units	No intervention (usual care)	22	3 mo
Iyer et al [83]	IG: 54; CG: 54	IG: 46.54 (10.31); CG: 43.93 (10.97)	IG: 54; CG: 54	IG: 38.07 (8.44); CG: 42.89 (11.49)	UCLA (20-item)	Online, self-care program	Wait-list control	4	4 wk
Karkosz et al [84]	IG: 40; CG: 41	IG: 45.44 (12.15); CG: 42.39 (13.25)	IG: 40; CG: 41	IG: 43.06 (13.38); CG: 41.49 (13.42)	UCLA (20-item)	Web-based and mobile therapy chatbot (CBT)	Active control (self-help book)	2	2 wk
Robinson et al [87]	IG: 17; CG: 17	IG: 36.44 (9.76); CG: 31.71 (9.50)	IG: 17; CG: 17	IG: 32.23 (9.92); CG: 33.93 (8.52)	UCLA (20-item)	Companion robot	Active control (standard activities)	12	12 wk
Shapira et al [88]	IG: 68; CG: 18	IG: 5.43 (2.0); CG: 6.11 (1.9)	IG: 64; CG: 18	IG: 4.80 (1.70); CG: 6.40 (1.70)	UCLA (3-item)	Online guided group sessions via Zoom (CBT)	Wait-list control	3.5	3.5 wk
Schwindenhammer [82]	IG: 40; CG: 40	IG: 17.44 (4.63); CG: 16.45 (4.43)	IG: 40; CG: 40	IG: 14.53 (4.23); CG: 14.13 (3.88)	UCLA (10-item)	Videoconferencing sessions	No intervention (usual care)	14	14 wk

^aIG: intervention group.

^bCG: control group.

^cUCLA: University of California, Los Angeles Loneliness Scale.

^dCBT: cognitive behavioral therapy.

The interventions were categorized into 3 distinct types based on their primary mechanism of action and content: (1) digital psychological interventions (n=4), which used technology to deliver structured therapeutic content (eg, cognitive behavioral therapy [CBT], mindfulness) or skills training. This category included a web-based self-help program [85], an online self-care program [83], a therapy chatbot [84], and online guided group CBT sessions [88]. (2) Information and Communication Technology for social connection (n=2), which focused on facilitating communication with others. This included videoconferencing with family members [82] and an online peer-support group [86]. (3) Social robots (n=1), represented by a companion seal robot designed to provide companionship [87].

The duration of interventions varied considerably, ranging from 2 weeks [84] to 22 weeks [86]. All studies assessed loneliness using the University of California, Los Angeles

Loneliness Scale, though different versions were used (eg, 3-item, 9-item, 10-item, or 20-item versions). Baseline loneliness scores indicated varying levels of loneliness across samples, from low to moderate to high.

Results of Syntheses

The meta-analysis demonstrated a small and statistically nonsignificant overall effect of interventions on loneliness (SMD=-0.21, 95% CI -0.59 to 0.17; P=.22), indicating that the interventions did not produce a reliable reduction in loneliness compared with control conditions (Table 3). The random-effects model was used to account for between-study heterogeneity, which was moderate (Q=13.68, df=6; P=.03; I²=57.44%, τ²=0.076; τ=0.276), suggesting that a substantial proportion of the variability reflects true differences among studies rather than sampling error.

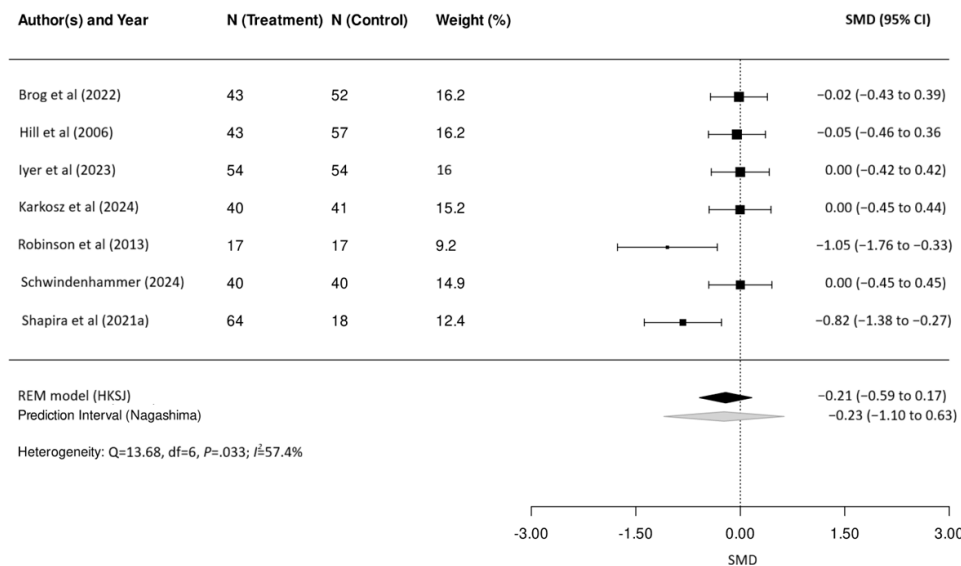
Table 3. Summary of random-effects meta-analysis results, showing standardized mean differences and statistical indicators for loneliness reduction in technology-based interventions vs control groups.

Study	Standardized mean difference	SE	Variance	95% CI	Z value	P value
Brog et al [85]	-0.019	0.208	0.043	-0.428 to 0.389	-0.093	.93
Hill et al [86]	-0.049	0.209	0.043	-0.457 to 0.360	-0.234	.82
Iyer et al [83]	0	0.212	0.045	-0.416 to 0.416	0.000	>.99
Karkosz et al [84]	-0.003	0.227	0.051	-0.448 to 0.441	-0.015	.99
Robinson et al [87]	-1.047	0.366	0.134	-1.764 to -0.330	-2.863	.004
Schwindenhammer [82]	0	0.231	0.053	-0.453 to 0.453	0.000	>.99
Shapira et al [88]	-0.824	0.282	0.080	-1.378 to -0.271	-2.919	.004

The estimated prediction interval ranges from -1.14 to 0.63, indicating that true effects in comparable future studies may vary from modest reductions to modest increases in loneliness (Figure 2). Given that the analysis included only 7 studies,

heterogeneity estimates and the prediction interval should be interpreted with caution. Further research is needed to better identify conditions under which loneliness interventions are effective.

Figure 2. Forest plot showing study-level standardized mean differences and 95% CIs for the effect of technology-based interventions on loneliness, synthesized using a Hartung-Knapp-Sidik-Jonkman adjusted random-effects model [82-88]. HKSJ: Hartung-Knapp-Sidik-Jonkman method; *I*²: *I*-squared heterogeneity statistic; Q: Cochran Q statistic; REML: restricted maximum-likelihood; SMD: standardized mean difference.



Risk of Bias

The summary of the risk of bias assessment is presented in Figure 3. Among the 7 included studies, 2 were judged to be at low risk of bias [85,88], 3 studies raised some concerns [83,84,87], and 2 studies were assessed as being at high risk of bias [82,86]. The high risk was primarily driven by domain 3 (missing outcome data) due to substantial differential attrition between the intervention and control groups [86] and

domain 4 (measurement of the outcome) due to the lack of blinding of the outcome assessor during data collection [82]. Most studies raised “some concerns” in domain 2 (deviations from intended interventions) and domain 4. This reflects the inherent nature of psychosocial and technological interventions, where blinding of participants is often not feasible, and the reliance on self-reported measures may introduce bias if the participants are aware of their group assignment.

Figure 3. Risk of bias summary: review authors’ judgments about each risk of bias item for each included study [82-88].

Study	Risk of bias domains					Overall
	D1	D2	D3	D4	D5	
Brog et al (2022)	+	-	+	+	+	+
Hill et al (2006)	-	X	X	-	-	X
Iyer et al (2023)	+	-	+	-	+	-
Karkosz et al (2024)	+	-	+	-	+	-
Robinson et al (2013)	-	+	+	-	+	-
S. Shapira et al (2021)	+	+	+	-	+	+
Schwindenhammer (2014)	X	X	+	X	+	X

Domains:
 D1: Bias arising from the randomization process.
 D2: Bias due to deviations from intended intervention.
 D3: Bias due to missing outcome data.
 D4: Bias in measurement of the outcome.
 D5: Bias in selection of the reported result.

Judgement
 High
 Some concerns
 Low

Reporting Biases

Visual inspection of the funnel plot and statistical testing were used to evaluate reporting bias. Egger regression test [89] showed statistically significant funnel plot asymmetry ($t_5=-5.47, P=.003$), indicating the presence of small-study effects (see Funnel Plot in Multimedia Appendix 3). However, Duval and Tweedie trim and fill method [90] did not identify any missing studies (0 missing studies), and no adjusted pooled effect was required. Accordingly, although funnel plot asymmetry was detected, the absence of imputed studies suggests that the asymmetry is likely driven by study heterogeneity (as indicated by the I^2 and wide prediction interval) rather than publication bias alone.

Sensitivity Analysis

A sensitivity analysis was conducted using a leave-one-out procedure in R, in which the pooled effect was recalculated after omitting each study in turn. The results showed that the pooled SMD remained small and nonsignificant across all iterations, and the direction of the effect did not change, indicating that no single study unduly influenced the overall findings.

Certainty of Evidence

The overall certainty of the evidence regarding the effect of technology-based interventions on loneliness was graded as very low. We downgraded the evidence by 1 level due to the risk of bias, as only 2 of the 7 included studies were rated as having a low risk of bias. We further downgraded the evidence due to inconsistencies caused by moderate heterogeneity and a wide prediction interval. Finally, we downgraded the rating due to imprecision, as the 95% CI for the pooled effect included both a noticeable benefit and no effect (crossing the 0 line). No publication bias was detected, but given the small number of studies, it could not be

completely ruled out. A detailed summary of the findings and certainty assessment is provided in Multimedia Appendix 4.

Discussion

Principal Findings

The primary objective of this meta-analysis was to evaluate the effectiveness of technology-based interventions in reducing loneliness. The results showed a small and statistically nonsignificant overall effect, indicating that the available evidence does not support a reliable reduction in loneliness attributable to technology-based interventions. Importantly, these findings must be interpreted with caution, as the certainty of the evidence was graded as very low. This low confidence stems primarily from limitations in the risk of bias domain and the observed inconsistency across studies. However, the moderate heterogeneity observed in the studies suggests that the effectiveness of these interventions varies across studies. The prediction interval indicated that true effects in comparable future studies could range from substantial reductions to moderate increases in loneliness, underscoring the influence of study-specific and contextual factors.

Because the number of available studies ($k=7$) did not permit reliable subgroup or moderator analyses, the present meta-analysis could not empirically evaluate how different types of technology-based interventions, such as digital psychological interventions or information and communication technology for social connection, vary in their effectiveness. However, previous research provides important insights into potential sources of variability. For example, some structured approaches, such as telephone-delivered behavioral activation or mindfulness interventions, have demonstrated beneficial effects compared with less structured social-contact interventions [89], where a telephone-based

behavioral activation or mindfulness intervention is effective compared with a control group that receives only a befriending intervention [89]. Moreover, classifying interventions by mechanism of action rather than just delivery mode offers insights for future research. For example, digital psychological interventions aiming to restructure maladaptive social cognition (eg, via CBT) have previously shown strong effects in face-to-face settings [28]. In our review, studies in this category [84,85,88] showed mixed results, potentially due to variations in guidance intensity. Research suggests that loneliness is a deeply personal experience shaped by social, cultural, and psychological factors [8,9]. As a result, intervention effectiveness is likely to depend on participant characteristics such as baseline loneliness, age, social resources, and psychological needs, although these factors could not be tested in the present analysis. Prior work suggests that digital mental health interventions are more effective when tailored to the needs and preferences of individuals [91], underscoring the importance of personalization rather than a uniform intervention model.

Despite the widespread interest in technology-based interventions, it is important to acknowledge the moderate proportion of heterogeneity in effects across studies. We observed an I^2 of 57%, indicating that more than half of the observed variance in effect sizes is due to true differences between studies rather than sampling error. This finding is consistent with prior observations that loneliness intervention outcomes are highly context-dependent [92-94]. However, the I^2 statistic does not reveal by how much the true effect varies across different populations. The prediction interval, in contrast, provides a direct estimate of this range. Previous reviews of loneliness interventions have also frequently reported substantial heterogeneity [28,95,96]. This variability may be due to differences in study design, with pre-post studies and nonrandomized comparative studies showing larger average effect sizes compared to randomized controlled studies [28]. The sources of heterogeneity could not be examined empirically in the present meta-analysis due to the small number of included studies; however, prior research suggests that study design, participant age, and digital literacy may influence intervention outcomes [97]. For example, younger participants often benefit more from digital programs that incorporate interactive or socially engaging elements, whereas older adults may experience additional barriers such as lower digital literacy or chronic social isolation that limit intervention impact [91,98-100]. Systematic reviews focusing specifically on older adults have reported minimal or no overall effects of digital interventions [75,101], underscoring the importance of population characteristics. Previous literature also indicates that intervention type, delivery format, and follow-up duration may influence loneliness outcomes [102-104], suggesting that program content and user engagement processes play important roles. Additionally, effects may vary by timing of assessment, with short-term improvements sometimes diminishing over longer follow-up periods [101,104]. Lastly, differences in loneliness measurement instruments and study quality cannot be overlooked. Most studies in the broader literature use the University of California, Los Angeles Loneliness Scale, but other

instruments, including single-item measures, are also used [105,106]. Meta-analytic evidence indicates that effect sizes can differ by instrument, partly due to scale sensitivity and range [107-110]. The use of a random-effects model helps account for some of this variability; however, the prediction interval we calculated in the present study was wide. This highlights a critical distinction: while the confidence interval suggests a nonsignificant average effect, the prediction interval implies that in specific future settings, a technology-based intervention might show only a negligible effect on loneliness or, conversely, a somewhat larger effect than the pooled estimate suggests.

A significant challenge encountered in this meta-analysis was the lack of essential data required for the computation of the effect size. Many studies failed to report sufficient statistical details, which considerably reduced the number of studies that could ultimately be included in the meta-analysis. This limitation was particularly striking given the large number of studies initially identified as suitable during the preliminary screening phase. Furthermore, some studies provided links to external repositories, where relevant data were supposed to be available. However, upon verification, these datasets were either missing or incomplete. This issue emphasizes the critical importance of comprehensive and transparent reporting in research. Consistent and standardized reporting of results, including the provision of complete datasets, would greatly enhance the feasibility and reliability of future meta-analyses and systematic reviews [63,111]. More rigorous data availability would allow for more robust conclusions with greater explanatory value, and so improve the overall quality and impact of evidence synthesis in the field.

Our meta-analysis found a small and statistically nonsignificant overall effect of technology-based interventions on loneliness, which should be interpreted in the context of prior research on loneliness interventions. Earlier meta-analyses in the field have generally reported small-to-moderate effects. For example, the landmark review by Masi et al [28], which examined various approaches such as social skills training, enhanced social support, and opportunities for social contact, found overall small reductions in loneliness, with the strongest effects observed for interventions addressing maladaptive social cognition [28,92]. In a more recent comprehensive review, Lasgaard et al [24] synthesized 128 studies and likewise reported a moderate loneliness reduction for randomized trials (SMD on the order of ~ -0.5). These earlier findings provide important context for interpreting our more modest pooled effect, and reinforce the conclusion that loneliness interventions in general tend to yield small-to-moderate benefits. Indeed, when focusing on digital interventions for older adults, recent meta-analyses have often found no significant effect: both Jin et al [75] and Shah et al [101] concluded that technology-based programs did not measurably reduce loneliness in older populations (pooled SMDs around 0 to -0.1 ; $P > .5$). An earlier analysis by Choi et al [112] had suggested that computer and internet training programs can be effective for older adults, but the evidence base at that time was limited to a few

small studies. Given that the present meta-analysis spans a broader age range, its findings contribute to a more nuanced picture of how digital interventions may operate across different demographic groups. The literature has yielded mixed results depending on population characteristics and intervention modality [24,28,103], and while such comparisons could not be undertaken in the present study, prior findings highlight the value of considering demographic and contextual factors when evaluating intervention outcomes. For instance, Eccles and Qualter [23] examined interventions among children and adolescents and found that while youth loneliness can be alleviated, effects varied greatly between studies, with substantial unexplained heterogeneity. Taken together, the broader evidence base suggests that intervention effectiveness may depend on factors such as target population and intervention format, though these influences could not be directly assessed in the present meta-analysis.

Strengths and Limitations

This meta-analysis adheres to PRISMA guidelines, ensuring methodological transparency and rigor. A comprehensive search strategy across multiple databases, supplemented by citation tracking, minimizes the risk of missing relevant studies. By incorporating prediction intervals alongside standard heterogeneity assessments, the analysis provides deeper insights into the variability of intervention effects. Furthermore, it contributes to the growing evidence base on technology-based interventions for loneliness, which is particularly relevant in the context of increasing workforce shortages in health care and social services.

However, the study has several limitations. First, the number of eligible studies that provided sufficient statistical information was small, which restricted the scope of the meta-analysis and limited the ability to conduct subgroup or moderator analyses. This constrains the conclusions that can be drawn regarding factors that may influence intervention effectiveness. Furthermore, several limitations of the review process should be acknowledged. Although we did not use formal language restrictions, we used English keywords for the search. This may have limited the identification of studies published in other languages. Additionally, our search strategy relied on major databases and citation tracking. We did not conduct a separate systematic search for gray literature, so some unpublished trials might have been missed. We also decided not to contact study authors for missing data because of the practical reasons explained in the *Methods*

section. This resulted in the exclusion of studies that could have been eligible if more information was available.

Implications

Technology is not a panacea for loneliness, yet it may serve as a useful tool when applied meaningfully and with consideration for the individual needs of the user. Communication technologies, such as video calls and messaging platforms, have been proposed as complements to interpersonal relationships and may help reduce feelings of loneliness in some contexts, particularly among older adults [113]. Similarly, social robots have emerged as nonpharmacological interventions capable of reducing both loneliness and depression [114]. However, a critical factor remains the manner in which these technologies are used. Active engagement in digital interactions has been shown to yield positive effects, whereas passive consumption of digital content may, conversely, exacerbate feelings of isolation [115]. Therefore, it is essential to consider users' age, technological literacy, and personal preferences when implementing technological solutions, to maximize their potential benefits and to avoid unintended increases in loneliness or social isolation.

Conclusions

This meta-analysis provides a rigorous synthesis of RCTs evaluating technology-based interventions for loneliness, including participants from adolescence to older adulthood. This study differs from existing reviews by focusing exclusively on randomized trials and the inclusion of a broader age demographic, whereas previous landmark studies often combined different study designs [28] or focused specifically on the older adult population [75]. Furthermore, the use of an RCT-only design, Hartung-Knapp adjustment, and prediction intervals offers a rigorous summary of the current evidence and highlights substantial variability in intervention outcomes. The small and nonsignificant pooled effect, together with moderate heterogeneity and a wide prediction interval, shows that results differ across studies and settings. These findings clarify current limitations in the evidence base and point to the need for better-designed and more transparently reported trials. In practice, technology-based approaches should be considered supportive tools that require careful tailoring to user characteristics and implementation contexts rather than universal solutions to loneliness.

Acknowledgments

The authors would like to express their gratitude to Marta Sykorova for her invaluable contribution to defining the search syntax, assisting in screening reports and studies, and extracting the necessary data. They also thank Kristyna Zivna for her efforts in identifying and extracting key statistical indicators essential for our analyses and meta-analytic calculations. The authors declare the use of generative artificial intelligence (AI) in the research and writing process. According to the GAIDeT taxonomy (2025), the following tasks were delegated to GAI tools under full human supervision:

Proofreading and editing

The GAI tool used was ChatGPT (GPT-5.1).

Responsibility for the final manuscript lies entirely with the authors.

GAI tools are not listed as authors and do not bear responsibility for the final outcomes.

Declaration submitted by: Collective responsibility

All authors declare that no other uses of large language models were made other than as stated.

Funding

The work was supported by ERDF/ESF project DigiWELL (CZ.02.01.01/00/22_008/0004583).

Data Availability

The study materials, data, and study code generated and/or analyzed during the current study are available in the Zenodo repository [116]. If any additional information regarding the data or analytic code is required, these are available from the corresponding author on reasonable request.

Authors' Contributions

Conceptualization: ZM, JS, PT

Data curation: MB, LN, OB

Formal analysis: JS, LN

Methodology: JS, MB, OB

Project administration: ZM, PT

Resources: JS, MB, OB

Supervision: ZM, PT

Writing – original draft: ZM, MB, JS, JH

Writing – review & editing: ZM, MB, JS, LN, JH, KM

All authors have read and approved the final version of the manuscript.

Conflicts of Interest

None declared.

Multimedia Appendix 1

Search syntaxes for database search.

[\[DOCX File \(Microsoft Word File\), 22 KB-Multimedia Appendix 1\]](#)

Multimedia Appendix 2

List of excluded studies.

[\[DOCX File \(Microsoft Word File\), 35 KB-Multimedia Appendix 2\]](#)

Multimedia Appendix 3

Funnel plot.

[\[PNG File \(Portable Network Graphics File\), 71 KB-Multimedia Appendix 3\]](#)

Multimedia Appendix 4

GRADE (Grading of Recommendations Assessment, Development, and Evaluation) summary of findings: technology-based interventions compared to control conditions for reducing loneliness.

[\[DOCX File \(Microsoft Word File\), 25 KB-Multimedia Appendix 4\]](#)

Checklist 1

PRISMA 2020 checklist, PRISMA 2020 for Abstracts checklist, and PRISMA-S checklist.

[\[PDF File \(Adobe File\), 367 KB-Checklist 1\]](#)

References

1. Hackett RA, Hudson JL, Chilcot J. Loneliness and type 2 diabetes incidence: findings from the English Longitudinal Study of Ageing. *Diabetologia*. Nov 2020;63(11):2329-2338. [doi: [10.1007/s00125-020-05258-6](https://doi.org/10.1007/s00125-020-05258-6)] [Medline: [32929525](https://pubmed.ncbi.nlm.nih.gov/32929525/)]
2. Leigh-Hunt N, Bagguley D, Bash K, et al. An overview of systematic reviews on the public health consequences of social isolation and loneliness. *Public Health (Fairfax)*. Nov 2017;152:157-171. [doi: [10.1016/j.puhe.2017.07.035](https://doi.org/10.1016/j.puhe.2017.07.035)] [Medline: [28915435](https://pubmed.ncbi.nlm.nih.gov/28915435/)]
3. Kobos E, Szewczyk A, Świątkowska T, Kryczka T, Sienkiewicz Z. Relationship between loneliness and blood glucose control in diabetes. *BMC Public Health*. Jul 20, 2020;20(1):1140. [doi: [10.1186/s12889-020-09241-z](https://doi.org/10.1186/s12889-020-09241-z)] [Medline: [32689971](https://pubmed.ncbi.nlm.nih.gov/32689971/)]
4. Wang X, Ma H, Li X, Heianza Y, Fonseca V, Qi L. Joint association of loneliness and traditional risk factor control and incident cardiovascular disease in diabetes patients. *Eur Heart J*. Jul 21, 2023;44(28):2583-2591. [doi: [10.1093/eurheartj/ehad306](https://doi.org/10.1093/eurheartj/ehad306)] [Medline: [37385629](https://pubmed.ncbi.nlm.nih.gov/37385629/)]
5. Zhou F, Deng L, Guo C, et al. Diabetes distress as mediators of loneliness and health promotion behaviour: a cross-sectional study. *BMJ Open*. Dec 28, 2023;13(12):e079674. [doi: [10.1136/bmjopen-2023-079674](https://doi.org/10.1136/bmjopen-2023-079674)] [Medline: [38154898](https://pubmed.ncbi.nlm.nih.gov/38154898/)]

6. Kuczmarski MF, Orsega-Smith E, Evans MK, Zonderman AB. The association of loneliness with diabetes is mediated by physical activity and BMI but not diet quality. *Nutrients*. Nov 25, 2023;15(23):4923. [doi: [10.3390/nu15234923](https://doi.org/10.3390/nu15234923)] [Medline: [38068781](https://pubmed.ncbi.nlm.nih.gov/38068781/)]
7. Richard A, Rohrmann S, Vandeleur CL, Schmid M, Barth J, Eichholzer M. Loneliness is adversely associated with physical and mental health and lifestyle factors: results from a Swiss national survey. *PLoS ONE*. 2017;12(7):e0181442. [doi: [10.1371/journal.pone.0181442](https://doi.org/10.1371/journal.pone.0181442)] [Medline: [28715478](https://pubmed.ncbi.nlm.nih.gov/28715478/)]
8. Solmi M, Veronese N, Galvano D, et al. Factors associated with loneliness: an umbrella review of observational studies. *J Affect Disord*. Jun 15, 2020;271:131-138. [doi: [10.1016/j.jad.2020.03.075](https://doi.org/10.1016/j.jad.2020.03.075)] [Medline: [32479308](https://pubmed.ncbi.nlm.nih.gov/32479308/)]
9. Beutel ME, Klein EM, Brähler E, et al. Loneliness in the general population: prevalence, determinants and relations to mental health. *BMC Psychiatry*. Mar 20, 2017;17(1):97. [doi: [10.1186/s12888-017-1262-x](https://doi.org/10.1186/s12888-017-1262-x)] [Medline: [28320380](https://pubmed.ncbi.nlm.nih.gov/28320380/)]
10. Erzen E, Çikrikci Ö. The effect of loneliness on depression: a meta-analysis. *Int J Soc Psychiatry*. Aug 2018;64(5):427-435. [doi: [10.1177/0020764018776349](https://doi.org/10.1177/0020764018776349)] [Medline: [29792097](https://pubmed.ncbi.nlm.nih.gov/29792097/)]
11. Wang J, Lloyd-Evans B, Marston L, Mann F, Ma R, Johnson S. Loneliness as a predictor of outcomes in mental disorders among people who have experienced a mental health crisis: a 4-month prospective study. *BMC Psychiatry*. May 20, 2020;20(1):249. [doi: [10.1186/s12888-020-02665-2](https://doi.org/10.1186/s12888-020-02665-2)] [Medline: [32434492](https://pubmed.ncbi.nlm.nih.gov/32434492/)]
12. Caple V, Maude P, Walter R, Ross A. An exploration of loneliness experienced by people living with mental illness and the impact on their recovery journey: an integrative review. *J Psychiatr Ment Health Nurs*. Dec 2023;30(6):1170-1191. [doi: [10.1111/jpm.12945](https://doi.org/10.1111/jpm.12945)] [Medline: [37382063](https://pubmed.ncbi.nlm.nih.gov/37382063/)]
13. Park C, Majeed A, Gill H, et al. The effect of loneliness on distinct health outcomes: a comprehensive review and meta-analysis. *Psychiatry Res*. Dec 2020;294:113514. [doi: [10.1016/j.psychres.2020.113514](https://doi.org/10.1016/j.psychres.2020.113514)] [Medline: [33130511](https://pubmed.ncbi.nlm.nih.gov/33130511/)]
14. Liang YY, Zhou M, He Y, et al. Observational and genetic evidence disagree on the association between loneliness and risk of multiple diseases. *Nat Hum Behav*. Nov 2024;8(11):2209-2221. [doi: [10.1038/s41562-024-01970-0](https://doi.org/10.1038/s41562-024-01970-0)] [Medline: [39284978](https://pubmed.ncbi.nlm.nih.gov/39284978/)]
15. Rico-Urbe LA, Caballero FF, Martín-María N, Cabello M, Ayuso-Mateos JL, Miret M. Association of loneliness with all-cause mortality: a meta-analysis. *PLoS ONE*. 2018;13(1):e0190033. [doi: [10.1371/journal.pone.0190033](https://doi.org/10.1371/journal.pone.0190033)] [Medline: [29300743](https://pubmed.ncbi.nlm.nih.gov/29300743/)]
16. Zhao J, Reese JB, Han X, Yabroff R. Associations of loneliness and mortality risk among cancer survivors in the United States. *J Clin Oncol*. Jun 1, 2023;41(16_suppl):6531-6531. [doi: [10.1200/JCO.2023.41.16_suppl.6531](https://doi.org/10.1200/JCO.2023.41.16_suppl.6531)] [Medline: [37034189](https://pubmed.ncbi.nlm.nih.gov/37034189/)]
17. Cacioppo JT, Hughes ME, Waite LJ, Hawkley LC, Thisted RA. Loneliness as a specific risk factor for depressive symptoms: cross-sectional and longitudinal analyses. *Psychol Aging*. Mar 2006;21(1):140-151. [doi: [10.1037/0882-7974.21.1.140](https://doi.org/10.1037/0882-7974.21.1.140)] [Medline: [16594799](https://pubmed.ncbi.nlm.nih.gov/16594799/)]
18. Baumeister RF, Leary MR. The need to belong: desire for interpersonal attachments as a fundamental human motivation. In: *Interpersonal Development*. Taylor & Francis; 2007:57-89. [doi: [10.4324/9781351153683-3](https://doi.org/10.4324/9781351153683-3)]
19. Hawkley LC. Loneliness and health. *Nat Rev Dis Primers*. Apr 14, 2022;8(1):22. [doi: [10.1038/s41572-022-00355-9](https://doi.org/10.1038/s41572-022-00355-9)] [Medline: [35422043](https://pubmed.ncbi.nlm.nih.gov/35422043/)]
20. Holt-Lunstad J, Smith TB, Baker M, Harris T, Stephenson D. Loneliness and social isolation as risk factors for mortality: a meta-analytic review. *Perspect Psychol Sci*. Mar 2015;10(2):227-237. [doi: [10.1177/1745691614568352](https://doi.org/10.1177/1745691614568352)] [Medline: [25910392](https://pubmed.ncbi.nlm.nih.gov/25910392/)]
21. Maes M, Vanhalst J. Loneliness as a double-edged sword: an adaptive function with maladaptive consequences. *Eur J Dev Psychol*. Mar 4, 2025;22(2):188-200. [doi: [10.1080/17405629.2024.2333584](https://doi.org/10.1080/17405629.2024.2333584)]
22. Lim MH, Holt-Lunstad J, Badcock JC. Loneliness: contemporary insights into causes, correlates, and consequences. *Soc Psychiatry Psychiatr Epidemiol*. Jul 2020;55(7):789-791. [doi: [10.1007/s00127-020-01891-z](https://doi.org/10.1007/s00127-020-01891-z)] [Medline: [32529475](https://pubmed.ncbi.nlm.nih.gov/32529475/)]
23. Eccles AM, Qualter P. Review: alleviating loneliness in young people - a meta-analysis of interventions. *Child Adolesc Ment Health*. Feb 2021;26(1):17-33. [doi: [10.1111/camh.12389](https://doi.org/10.1111/camh.12389)] [Medline: [32406165](https://pubmed.ncbi.nlm.nih.gov/32406165/)]
24. Lasgaard M, Løvschall C, Qualter P, et al. Are loneliness interventions effective in reducing loneliness? A meta-analytic review of 128 studies. *Eur J Public Health*. Oct 21, 2022;32(Supplement_3):ckac129.266. [doi: [10.1093/eurpub/ckac129.266](https://doi.org/10.1093/eurpub/ckac129.266)]
25. Osborn T, Weatherburn P, French RS. Interventions to address loneliness and social isolation in young people: a systematic review of the evidence on acceptability and effectiveness. *J Adolesc*. Dec 2021;93:53-79. [doi: [10.1016/j.adolescence.2021.09.007](https://doi.org/10.1016/j.adolescence.2021.09.007)] [Medline: [34662802](https://pubmed.ncbi.nlm.nih.gov/34662802/)]
26. Crowe CL, Liu L, Bagnarol N, Fried LP. Loneliness prevention and the role of the public health system. *Perspect Public Health*. Jan 2024;144(1):31-38. [doi: [10.1177/17579139221106579](https://doi.org/10.1177/17579139221106579)] [Medline: [35815809](https://pubmed.ncbi.nlm.nih.gov/35815809/)]
27. Patil U, Braun KL. Interventions for loneliness in older adults: a systematic review of reviews. *Front Public Health*. 2024;12:1427605. [doi: [10.3389/fpubh.2024.1427605](https://doi.org/10.3389/fpubh.2024.1427605)] [Medline: [39091527](https://pubmed.ncbi.nlm.nih.gov/39091527/)]

28. Masi CM, Chen HY, Hawkey LC, Cacioppo JT. A meta-analysis of interventions to reduce loneliness. *Pers Soc Psychol Rev.* Aug 2011;15(3):219-266. [doi: [10.1177/1088868310377394](https://doi.org/10.1177/1088868310377394)] [Medline: [20716644](https://pubmed.ncbi.nlm.nih.gov/20716644/)]
29. Peplau LA, Perlman D. *Loneliness: A Sourcebook of Current Theory, Research, and Therapy.* John Wiley & Sons; 1982. ISBN: 978-0471080282
30. Norlin J, McKee KJ, Lennartsson C, Dahlberg L. Quantity and quality of social relationships and their associations with loneliness in older adults. *Aging Ment Health.* Jul 2025;29(7):1198-1208. [doi: [10.1080/13607863.2025.2460068](https://doi.org/10.1080/13607863.2025.2460068)] [Medline: [39951254](https://pubmed.ncbi.nlm.nih.gov/39951254/)]
31. Weiss RS. *Loneliness: The Experience of Emotional and Social Isolation.* MIT press; 1973. URL: <https://psycnet.apa.org/record/1974-22306-000> [Accessed 2026-04-14]
32. Bolmsjö I, Tengland PA, Rängård M. Existential loneliness: an attempt at an analysis of the concept and the phenomenon. *Nurs Ethics.* Aug 2019;26(5):1310-1325. [doi: [10.1177/0969733017748480](https://doi.org/10.1177/0969733017748480)] [Medline: [29471724](https://pubmed.ncbi.nlm.nih.gov/29471724/)]
33. Cacioppo S, Grippo AJ, London S, Goossens L, Cacioppo JT. Loneliness: clinical import and interventions. *Perspect Psychol Sci.* Mar 2015;10(2):238-249. [doi: [10.1177/1745691615570616](https://doi.org/10.1177/1745691615570616)] [Medline: [25866548](https://pubmed.ncbi.nlm.nih.gov/25866548/)]
34. Hawkey LC, Browne MW, Cacioppo JT. How can I connect with thee? Let me count the ways. *Psychol Sci.* Oct 2005;16(10):798-804. [doi: [10.1111/j.1467-9280.2005.01617.x](https://doi.org/10.1111/j.1467-9280.2005.01617.x)] [Medline: [16181443](https://pubmed.ncbi.nlm.nih.gov/16181443/)]
35. Hawkey LC, Gu Y, Luo YJ, Cacioppo JT. The mental representation of social connections: generalizability extended to Beijing adults. *PLoS ONE.* 2012;7(9):e44065. [doi: [10.1371/journal.pone.0044065](https://doi.org/10.1371/journal.pone.0044065)] [Medline: [23028486](https://pubmed.ncbi.nlm.nih.gov/23028486/)]
36. Malli MA, Ryan S, Maddison J, Kharicha K. Experiences and meaning of loneliness beyond age and group identity. *Sociol Health Illn.* Jan 2023;45(1):70-89. [doi: [10.1111/1467-9566.13539](https://doi.org/10.1111/1467-9566.13539)] [Medline: [36073866](https://pubmed.ncbi.nlm.nih.gov/36073866/)]
37. Yang K, Victor C. Age and loneliness in 25 European nations. *Ageing Soc.* Nov 2011;31(8):1368-1388. [doi: [10.1017/S0144686X1000139X](https://doi.org/10.1017/S0144686X1000139X)]
38. Luhmann M, Hawkey LC. Age differences in loneliness from late adolescence to oldest old age. *Dev Psychol.* Jun 2016;52(6):943-959. [doi: [10.1037/dev0000117](https://doi.org/10.1037/dev0000117)] [Medline: [27148782](https://pubmed.ncbi.nlm.nih.gov/27148782/)]
39. Mund M, Lütke O, Neyer FJ. Owner of a lonely heart: the stability of loneliness across the life span. *J Pers Soc Psychol.* 2020;119(2):497-516. [doi: [10.1037/pspp0000262.supp](https://doi.org/10.1037/pspp0000262.supp)]
40. Ong AD, Uchino BN, Wethington E. Loneliness and health in older adults: a mini-review and synthesis. *Gerontology.* 2016;62(4):443-449. [doi: [10.1159/000441651](https://doi.org/10.1159/000441651)] [Medline: [26539997](https://pubmed.ncbi.nlm.nih.gov/26539997/)]
41. von Soest T, Luhmann M, Hansen T, Gerstorf D. Development of loneliness in midlife and old age: its nature and correlates. *J Pers Soc Psychol.* Feb 2020;118(2):388-406. [doi: [10.1037/pspp0000219](https://doi.org/10.1037/pspp0000219)] [Medline: [30284871](https://pubmed.ncbi.nlm.nih.gov/30284871/)]
42. Igami K, Hosozawa M, Ikeda A, Bann D, Shimizu T, Iso H. Adolescent loneliness in 70 countries across Africa, America, and Asia: a comparison of prevalence and correlates. *J Adolesc Health.* Jun 2023;72(6):906-913. [doi: [10.1016/j.jadohealth.2022.12.029](https://doi.org/10.1016/j.jadohealth.2022.12.029)] [Medline: [36775749](https://pubmed.ncbi.nlm.nih.gov/36775749/)]
43. Su Y, Rao W, Li M, Caron G, D'Arcy C, Meng X. Prevalence of loneliness and social isolation among older adults during the COVID-19 pandemic: a systematic review and meta-analysis. *Int Psychogeriatr.* May 2023;35(5):229-241. [doi: [10.1017/S1041610222000199](https://doi.org/10.1017/S1041610222000199)] [Medline: [35357280](https://pubmed.ncbi.nlm.nih.gov/35357280/)]
44. Anderson GO, Thayer CE. *Loneliness and social connections: a national survey of adults 45 and older.* AARP Foundation; 2018. URL: <https://www.icmha.org/wp-content/uploads/2020/02/AARP-740C-Loneliness-Report-v3-with-LOGO-TAG.pdf> [Accessed 2026-04-14]
45. Cigna U.S. *loneliness index: survey of 20,000 Americans examining behaviors driving loneliness in the United States.* Cigna; 2018. URL: <https://www.cigna.com/assets/docs/newsroom/loneliness-survey-2018-full-report.pdf> [Accessed 2026-04-14]
46. *Loneliness and the workplace: 2020 U.S. report.* Cigna; 2020. URL: <https://www.cigna.com/static/www-cigna-com/docs/about-us/newsroom/studies-and-reports/combating-loneliness/cigna-2020-loneliness-report.pdf> [Accessed 2026-04-14]
47. Stegen H, Duppen D, Savieri P, et al. Loneliness prevalence of community-dwelling older adults and the impact of the mode of measurement, data collection, and country: a systematic review and meta-analysis. *Int Psychogeriatr.* Sep 2024;36(9):747-761. [doi: [10.1017/S1041610224000425](https://doi.org/10.1017/S1041610224000425)] [Medline: [38525677](https://pubmed.ncbi.nlm.nih.gov/38525677/)]
48. Yang K. *Loneliness: A Social Problem.* Routledge; 2019. [doi: [10.4324/9781315148410](https://doi.org/10.4324/9781315148410)]
49. Graham EK, Beck ED, Jackson K, et al. Do we become more lonely with age? A coordinated data analysis of nine longitudinal studies. *Psychol Sci.* Jun 2024;35(6):579-596. [doi: [10.1177/09567976241242037](https://doi.org/10.1177/09567976241242037)] [Medline: [38687352](https://pubmed.ncbi.nlm.nih.gov/38687352/)]
50. Lay-Yee R, Campbell D, Milne B. Social attitudes and activities associated with loneliness: findings from a New Zealand national survey of the adult population. *Health Soc Care Community.* May 2022;30(3):1120-1132. [doi: [10.1111/hsc.13351](https://doi.org/10.1111/hsc.13351)] [Medline: [33765362](https://pubmed.ncbi.nlm.nih.gov/33765362/)]

51. Nyqvist F, Victor CR, Forsman AK, Cattani M. The association between social capital and loneliness in different age groups: a population-based study in Western Finland. *BMC Public Health*. Jul 11, 2016;16:542. [doi: [10.1186/s12889-016-3248-x](https://doi.org/10.1186/s12889-016-3248-x)] [Medline: [27400659](https://pubmed.ncbi.nlm.nih.gov/27400659/)]
52. Hickin N, Käll A, Shafran R, Sutcliffe S, Manzotti G, Langan D. The effectiveness of psychological interventions for loneliness: a systematic review and meta-analysis. *Clin Psychol Rev*. Aug 2021;88:102066. [doi: [10.1016/j.cpr.2021.102066](https://doi.org/10.1016/j.cpr.2021.102066)] [Medline: [34339939](https://pubmed.ncbi.nlm.nih.gov/34339939/)]
53. Baarek J, d'Hombres B, Tintori G. Loneliness in Europe before and during the COVID-19 pandemic. *Health Policy*. Nov 2022;126(11):1124-1129. [doi: [10.1016/j.healthpol.2022.09.002](https://doi.org/10.1016/j.healthpol.2022.09.002)] [Medline: [36182348](https://pubmed.ncbi.nlm.nih.gov/36182348/)]
54. Banerjee D, Rai M. Social isolation in Covid-19: the impact of loneliness. *Int J Soc Psychiatry*. Sep 2020;66(6):525-527. [doi: [10.1177/0020764020922269](https://doi.org/10.1177/0020764020922269)] [Medline: [32349580](https://pubmed.ncbi.nlm.nih.gov/32349580/)]
55. Hwang TJ, Rabheru K, Peisah C, Reichman W, Ikeda M. Loneliness and social isolation during the COVID-19 pandemic. *Int Psychogeriatr*. Oct 2020;32(10):1217-1220. [doi: [10.1017/S1041610220000988](https://doi.org/10.1017/S1041610220000988)] [Medline: [32450943](https://pubmed.ncbi.nlm.nih.gov/32450943/)]
56. Patel SS, Clark-Ginsberg A. Incorporating issues of elderly loneliness into the Coronavirus Disease-2019 public health response. *Disaster Med Public Health Prep*. Jun 2020;14(3):e13-e14. [doi: [10.1017/dmp.2020.145](https://doi.org/10.1017/dmp.2020.145)] [Medline: [32379016](https://pubmed.ncbi.nlm.nih.gov/32379016/)]
57. Hajek A, Sutin AR, Posi G, et al. Chronic loneliness and chronic social isolation among older adults. A systematic review, meta-analysis and meta-regression. *Aging Ment Health*. Feb 2025;29(2):185-200. [doi: [10.1080/13607863.2024.2385448](https://doi.org/10.1080/13607863.2024.2385448)] [Medline: [39126212](https://pubmed.ncbi.nlm.nih.gov/39126212/)]
58. Shah SGS, Noguera D, van Woerden HC, Kiparoglou V. Evaluation of the effectiveness of digital technology interventions to reduce loneliness in older adults: systematic review and meta-analysis. *J Med Internet Res*. 2021;23(6):e24712. [doi: [10.2196/24712](https://doi.org/10.2196/24712)] [Medline: [34085942](https://pubmed.ncbi.nlm.nih.gov/34085942/)]
59. Bentley SV, Haslam C, Haslam SA, Jetten J, Larwood J, La Rue CJ. GROUPS 2 CONNECT: an online activity to maintain social connection and well-being during COVID-19. *Appl Psychol Health Well Being*. Nov 2022;14(4):1189-1210. [doi: [10.1111/aphw.12330](https://doi.org/10.1111/aphw.12330)] [Medline: [34958172](https://pubmed.ncbi.nlm.nih.gov/34958172/)]
60. Kenyon K, Kinakh V, Harrison J. Social virtual reality helps to reduce feelings of loneliness and social anxiety during the Covid-19 pandemic. *Sci Rep*. Nov 7, 2023;13(1):19282. [doi: [10.1038/s41598-023-46494-1](https://doi.org/10.1038/s41598-023-46494-1)] [Medline: [37935718](https://pubmed.ncbi.nlm.nih.gov/37935718/)]
61. Sandora J, Meier Z, Tavel P, Buchtova M, Sykorova M. The effectiveness of technology-based interventions on reduction of loneliness: systematic review and meta-analysis. *PROSPERO*; 2024. URL: <https://www.crd.york.ac.uk/PROSPERO/view/CRD42024505117> [Accessed 2026-04-14]
62. Moher D, Liberati A, Tetzlaff J, Altman DG, PRISMA Group. Preferred reporting items for systematic reviews and meta-analyses: the PRISMA statement. *Ann Intern Med*. Aug 18, 2009;151(4):264-269. [doi: [10.7326/0003-4819-151-4-200908180-00135](https://doi.org/10.7326/0003-4819-151-4-200908180-00135)] [Medline: [19622511](https://pubmed.ncbi.nlm.nih.gov/19622511/)]
63. Page MJ, McKenzie JE, Bossuyt PM, et al. The PRISMA 2020 statement: an updated guideline for reporting systematic reviews. *BMJ*. Mar 29, 2021;372:n71. [doi: [10.1136/bmj.n71](https://doi.org/10.1136/bmj.n71)] [Medline: [33782057](https://pubmed.ncbi.nlm.nih.gov/33782057/)]
64. Rethlefsen ML, Kirtley S, Waffenschmidt S, et al. PRISMA-S: an extension to the PRISMA statement for reporting literature searches in systematic reviews. *Syst Rev*. Jan 26, 2021;10(1):39. [doi: [10.1186/s13643-020-01542-z](https://doi.org/10.1186/s13643-020-01542-z)] [Medline: [33499930](https://pubmed.ncbi.nlm.nih.gov/33499930/)]
65. Zotero. 2024. URL: <https://www.zotero.org/> [Accessed 2026-04-14]
66. Borenstein M, Hedges LV, Higgins JPT, Rothstein HR. *Introduction to Meta-Analysis*. John Wiley & Sons; 2009. [doi: [10.1002/9780470743386](https://doi.org/10.1002/9780470743386)]
67. Cohen J. *Statistical Power Analysis for the Behavioral Sciences*. 2nd ed. Routledge; 1988. [doi: [10.4324/9780203771587](https://doi.org/10.4324/9780203771587)]
68. Int'Hout J, Ioannidis JPA, Borm GF. The Hartung-Knapp-Sidik-Jonkman method for random effects meta-analysis is straightforward and considerably outperforms the standard DerSimonian-Laird method. *BMC Med Res Methodol*. Feb 18, 2014;14:25. [doi: [10.1186/1471-2288-14-25](https://doi.org/10.1186/1471-2288-14-25)] [Medline: [24548571](https://pubmed.ncbi.nlm.nih.gov/24548571/)]
69. Nagashima K, Noma H, Furukawa TA. Prediction intervals for random-effects meta-analysis: a confidence distribution approach. *Stat Methods Med Res*. Jun 2019;28(6):1689-1702. [doi: [10.1177/0962280218773520](https://doi.org/10.1177/0962280218773520)] [Medline: [29745296](https://pubmed.ncbi.nlm.nih.gov/29745296/)]
70. Sterne JAC, Savović J, Page MJ, et al. RoB 2: a revised tool for assessing risk of bias in randomised trials. *BMJ*. Aug 28, 2019;366:14898. [doi: [10.1136/bmj.l4898](https://doi.org/10.1136/bmj.l4898)] [Medline: [31462531](https://pubmed.ncbi.nlm.nih.gov/31462531/)]
71. McGuinness LA, Higgins JPT. Risk-of-bias VISualization (robvis): an R package and Shiny web app for visualizing risk-of-bias assessments. *Res Synth Methods*. Jan 2021;12(1):55-61. [doi: [10.1002/jrsm.1411](https://doi.org/10.1002/jrsm.1411)] [Medline: [32336025](https://pubmed.ncbi.nlm.nih.gov/32336025/)]
72. Guyatt GH, Oxman AD, Vist GE, et al. GRADE: an emerging consensus on rating quality of evidence and strength of recommendations. *BMJ*. Apr 26, 2008;336(7650):924-926. [doi: [10.1136/bmj.39489.470347.AD](https://doi.org/10.1136/bmj.39489.470347.AD)] [Medline: [18436948](https://pubmed.ncbi.nlm.nih.gov/18436948/)]
73. Sun L. Social media usage and students' social anxiety, loneliness and well-being: does digital mindfulness-based intervention effectively work? *BMC Psychol*. Oct 31, 2023;11(1):362. [doi: [10.1186/s40359-023-01398-7](https://doi.org/10.1186/s40359-023-01398-7)] [Medline: [37904182](https://pubmed.ncbi.nlm.nih.gov/37904182/)]

74. Sun L. Retraction note: social media usage and students' social anxiety, loneliness and well-being: does digital mindfulness-based intervention effectively work? *BMC Psychol*. May 7, 2025;13(1):480. [doi: [10.1186/s40359-025-02810-0](https://doi.org/10.1186/s40359-025-02810-0)] [Medline: [40336136](https://pubmed.ncbi.nlm.nih.gov/40336136/)]
75. Jin W, Liu Y, Yuan S, Bai R, Li X, Bai Z. The effectiveness of technology-based interventions for reducing loneliness in older adults: a systematic review and meta-analysis of randomized controlled trials. *Front Psychol*. 2021;12:711030. [doi: [10.3389/fpsyg.2021.711030](https://doi.org/10.3389/fpsyg.2021.711030)] [Medline: [34955948](https://pubmed.ncbi.nlm.nih.gov/34955948/)]
76. Shapira N, Barak A, Gal I. Promoting older adults' well-being through internet training and use. *Aging Ment Health*. Sep 2007;11(5):477-484. [doi: [10.1080/13607860601086546](https://doi.org/10.1080/13607860601086546)] [Medline: [17882585](https://pubmed.ncbi.nlm.nih.gov/17882585/)]
77. Hind D, Mountain G, Gossage-Worrall R, et al. Putting Life in Years (PLINY): a randomised controlled trial and mixed-methods process evaluation of a telephone friendship intervention to improve mental well-being in independently living older people. *Public Health Res*. 2014;2(7):1-222. [doi: [10.3310/phr02070](https://doi.org/10.3310/phr02070)] [Medline: [25642518](https://pubmed.ncbi.nlm.nih.gov/25642518/)]
78. Tsai HH, Tsai YF, Wang HH, Chang YC, Chu HH. Videoconference program enhances social support, loneliness, and depressive status of elderly nursing home residents. *Aging Ment Health*. Nov 2010;14(8):947-954. [doi: [10.1080/13607863.2010.501057](https://doi.org/10.1080/13607863.2010.501057)] [Medline: [21069600](https://pubmed.ncbi.nlm.nih.gov/21069600/)]
79. Tsai HH, Cheng CY, Shieh WY, Chang YC. Effects of a smartphone-based videoconferencing program for older nursing home residents on depression, loneliness, and quality of life: a quasi-experimental study. *BMC Geriatr*. Jan 28, 2020;20(1):27. [doi: [10.1186/s12877-020-1426-2](https://doi.org/10.1186/s12877-020-1426-2)] [Medline: [31992217](https://pubmed.ncbi.nlm.nih.gov/31992217/)]
80. Tsai HH, Tsai YF. Changes in depressive symptoms, social support, and loneliness over 1 year after a minimum 3-month videoconference program for older nursing home residents. *J Med Internet Res*. Nov 15, 2011;13(4):e93. [doi: [10.2196/jmir.1678](https://doi.org/10.2196/jmir.1678)] [Medline: [22086660](https://pubmed.ncbi.nlm.nih.gov/22086660/)]
81. Slegers K, van Boxtel MPJ, Jolles J. Effects of computer training and internet usage on the well-being and quality of life of older adults: a randomized, controlled study. *J Gerontol B Psychol Sci Soc Sci*. May 2008;63(3):176-184. [doi: [10.1093/geronb/63.3.p176](https://doi.org/10.1093/geronb/63.3.p176)] [Medline: [18559683](https://pubmed.ncbi.nlm.nih.gov/18559683/)]
82. Schwindenhammer TM. Videoconferencing intervention for depressive symptoms and loneliness in nursing home elders [Master's Thesis]. Illinois State University; 2014. URL: <https://www.proquest.com/docview/1548716414/abstract/452A58D6944F4846PQ/1> [Accessed 2026-04-14]
83. Iyer RB, Vadlapudi S, Iyer L, et al. Impact of the Heartfulness program on loneliness in high schoolers: randomized survey study. *Appl Psychol Health Well Being*. Feb 2023;15(1):66-79. [doi: [10.1111/aphw.12360](https://doi.org/10.1111/aphw.12360)] [Medline: [35384302](https://pubmed.ncbi.nlm.nih.gov/35384302/)]
84. Karkosz S, Szymański R, Sanna K, Michałowski J. Effectiveness of a web-based and mobile therapy chatbot on anxiety and depressive symptoms in subclinical young adults: randomized controlled trial. *JMIR Form Res*. Mar 20, 2024;8(1):e47960. [doi: [10.2196/47960](https://doi.org/10.2196/47960)] [Medline: [38506892](https://pubmed.ncbi.nlm.nih.gov/38506892/)]
85. Brog NA, Hegy JK, Berger T, Znoj H. Effects of an internet-based self-help intervention for psychological distress due to COVID-19: results of a randomized controlled trial. *Internet Interv*. Mar 2022;27:100492. [doi: [10.1016/j.invent.2021.100492](https://doi.org/10.1016/j.invent.2021.100492)] [Medline: [34956841](https://pubmed.ncbi.nlm.nih.gov/34956841/)]
86. Hill W, Weinert C, Cudney S. Influence of a computer intervention on the psychological status of chronically ill rural women: preliminary results. *Nurs Res*. 2006;55(1):34-42. [doi: [10.1097/00006199-200601000-00005](https://doi.org/10.1097/00006199-200601000-00005)] [Medline: [16439927](https://pubmed.ncbi.nlm.nih.gov/16439927/)]
87. Robinson H, Macdonald B, Kerse N, Broadbent E. The psychosocial effects of a companion robot: a randomized controlled trial. *J Am Med Dir Assoc*. Sep 2013;14(9):661-667. [doi: [10.1016/j.jamda.2013.02.007](https://doi.org/10.1016/j.jamda.2013.02.007)] [Medline: [23545466](https://pubmed.ncbi.nlm.nih.gov/23545466/)]
88. Shapira S, Yeshua-Katz D, Cohn-Schwartz E, Aharonson-Daniel L, Sarid O, Clarfield AM. A pilot randomized controlled trial of a group intervention via Zoom to relieve loneliness and depressive symptoms among older persons during the COVID-19 outbreak. *Internet Interv*. Apr 2021;24:100368. [doi: [10.1016/j.invent.2021.100368](https://doi.org/10.1016/j.invent.2021.100368)] [Medline: [33527072](https://pubmed.ncbi.nlm.nih.gov/33527072/)]
89. Jiang D, Yeung D, Choi N, et al. Effectiveness of teledelivered behavioral activation and mindfulness interventions in reducing loneliness. *Innov Aging*. Dec 21, 2023;7(Supplement_1):844. [doi: [10.1093/geroni/igad104.2722](https://doi.org/10.1093/geroni/igad104.2722)]
90. Duval S, Tweedie R. Trim and fill: a simple funnel-plot-based method of testing and adjusting for publication bias in meta-analysis. *Biometrics*. Jun 2000;56(2):455-463. [doi: [10.1111/j.0006-341x.2000.00455.x](https://doi.org/10.1111/j.0006-341x.2000.00455.x)] [Medline: [10877304](https://pubmed.ncbi.nlm.nih.gov/10877304/)]
91. Burke L. Interventions to reduce loneliness in children and adolescents: a systematic review and meta-analysis. *Eur J Public Health*. Nov 1, 2024;34(Supplement_3):ckae144. [doi: [10.1093/eurpub/ckae144.056](https://doi.org/10.1093/eurpub/ckae144.056)]
92. Beckers A, Buecker S, Casabianca EJ, Nurminen M. Effectiveness of interventions tackling loneliness. European Commission; 2022. URL: https://www.researchgate.net/publication/366055118_Effectiveness_of_interventions_tackling_loneliness [Accessed 2026-04-14]
93. Cruwys T, Haslam C, Haslam SA, Dingle GA. Misery loves company: predictors of treatment response to a loneliness intervention. *Psychother Res*. Jun 2023;33(5):608-624. [doi: [10.1080/10503307.2022.2143300](https://doi.org/10.1080/10503307.2022.2143300)] [Medline: [36404283](https://pubmed.ncbi.nlm.nih.gov/36404283/)]

94. Ratcliffe J, Matcham F, Molteni E, et al. Conceptualising the context and mechanisms for tackling loneliness in older adults through interventions: a critical interpretive synthesis. medRxiv. Preprint posted online on Jan 6, 2025. [doi: [10.1101/2025.01.06.25320003](https://doi.org/10.1101/2025.01.06.25320003)]
95. Fu Z, Yan M, Meng C. The effectiveness of remote delivered intervention for loneliness reduction in older adults: a systematic review and meta-analysis. *Front Psychol*. 2022;13:935544. [doi: [10.3389/fpsyg.2022.935544](https://doi.org/10.3389/fpsyg.2022.935544)] [Medline: [35967719](https://pubmed.ncbi.nlm.nih.gov/35967719/)]
96. Grillich L, Titscher V, Klingenstein P, et al. The effectiveness of interventions to prevent loneliness and social isolation in the community-dwelling and old population: an overview of systematic reviews and meta-analysis. *Eur J Public Health*. Apr 1, 2023;33(2):235-241. [doi: [10.1093/eurpub/ckad006](https://doi.org/10.1093/eurpub/ckad006)] [Medline: [36893335](https://pubmed.ncbi.nlm.nih.gov/36893335/)]
97. van der Vaart R, van Driel D, Pronk K, et al. The role of age, education, and digital health literacy in the usability of internet-based cognitive behavioral therapy for chronic pain: mixed methods study. *JMIR Form Res*. Nov 21, 2019;3(4):e12883. [doi: [10.2196/12883](https://doi.org/10.2196/12883)] [Medline: [31750839](https://pubmed.ncbi.nlm.nih.gov/31750839/)]
98. Cohen-Mansfield J, Perach R. Interventions for alleviating loneliness among older persons: a critical review. *Am J Health Promot*. 2015;29(3):e109-25. [doi: [10.4278/ajhp.130418-LIT-182](https://doi.org/10.4278/ajhp.130418-LIT-182)] [Medline: [24575725](https://pubmed.ncbi.nlm.nih.gov/24575725/)]
99. Eager S, Johnson S, Pitman A, Uribe M, Qualter P, Pearce E. Young people's views on the acceptability and feasibility of loneliness interventions for their age group. *BMC Psychiatry*. Apr 23, 2024;24(1):308. [doi: [10.1186/s12888-024-05751-x](https://doi.org/10.1186/s12888-024-05751-x)] [Medline: [38654301](https://pubmed.ncbi.nlm.nih.gov/38654301/)]
100. Shekelle PG, Mlake-Lye IM, Begashaw MM, et al. Interventions to reduce loneliness in community-living older adults: a systematic review and meta-analysis. *J Gen Intern Med*. May 2024;39(6):1015-1028. [doi: [10.1007/s11606-023-08517-5](https://doi.org/10.1007/s11606-023-08517-5)] [Medline: [38200279](https://pubmed.ncbi.nlm.nih.gov/38200279/)]
101. Shah SGS, Nogueras D, van Woerden H, Kiparoglou V. Effectiveness of digital technology interventions to reduce loneliness in adults: a protocol for a systematic review and meta-analysis. *BMJ Open*. Sep 27, 2019;9(9):e032455. [doi: [10.1136/bmjopen-2019-032455](https://doi.org/10.1136/bmjopen-2019-032455)] [Medline: [31562164](https://pubmed.ncbi.nlm.nih.gov/31562164/)]
102. Centre for Reviews and Dissemination. University of York. 2014. URL: <https://www.york.ac.uk/crd> [Accessed 2026-04-14]
103. Morrish N, Choudhury S, Medina-Lara A. What works in interventions targeting loneliness: a systematic review of intervention characteristics. *BMC Public Health*. Nov 9, 2023;23(1):2214. [doi: [10.1186/s12889-023-17097-2](https://doi.org/10.1186/s12889-023-17097-2)] [Medline: [37946186](https://pubmed.ncbi.nlm.nih.gov/37946186/)]
104. Blodgett JM, Tiley K, Harkness F, Musella M. What works to reduce loneliness: a rapid systematic review of 101 interventions. *J Public Health Policy*. Jun 2025;46(2):245-268. [doi: [10.1057/s41271-025-00561-1](https://doi.org/10.1057/s41271-025-00561-1)] [Medline: [40050496](https://pubmed.ncbi.nlm.nih.gov/40050496/)]
105. Das A, Padala KP, Crawford CG, et al. A systematic review of loneliness and social isolation scales used in epidemics and pandemics. *Psychiatry Res*. Dec 2021;306:114217. [doi: [10.1016/j.psychres.2021.114217](https://doi.org/10.1016/j.psychres.2021.114217)] [Medline: [34644661](https://pubmed.ncbi.nlm.nih.gov/34644661/)]
106. Maes M, Qualter P, Lodder GMA, Mund M. How (not) to measure loneliness: a review of the eight most commonly used scales. *Int J Environ Res Public Health*. Aug 30, 2022;19(17):10816. [doi: [10.3390/ijerph191710816](https://doi.org/10.3390/ijerph191710816)] [Medline: [36078540](https://pubmed.ncbi.nlm.nih.gov/36078540/)]
107. Hudiyana J, Lincoln TM, Hartanto S, et al. How universal is a construct of loneliness? Measurement invariance of the UCLA loneliness scale in Indonesia, Germany, and the United States. *Assessment*. Dec 2022;29(8):1795-1805. [doi: [10.1177/10731911211034564](https://doi.org/10.1177/10731911211034564)] [Medline: [34301150](https://pubmed.ncbi.nlm.nih.gov/34301150/)]
108. Manera KE, Smith BJ, Owen KB, Phongsavan P, Lim MH. Psychometric assessment of scales for measuring loneliness and social isolation: an analysis of the household, income and labour dynamics in Australia (HILDA) survey. *Health Qual Life Outcomes*. Mar 5, 2022;20(1):40. [doi: [10.1186/s12955-022-01946-6](https://doi.org/10.1186/s12955-022-01946-6)] [Medline: [35248075](https://pubmed.ncbi.nlm.nih.gov/35248075/)]
109. Schoenmakers EC, Lasgaard M, McHugh Power J. Guidelines for evaluating and reporting social isolation and loneliness interventions. *J Health Psychol*. Feb 2025;30(2):338-352. [doi: [10.1177/13591053241238127](https://doi.org/10.1177/13591053241238127)] [Medline: [38527950](https://pubmed.ncbi.nlm.nih.gov/38527950/)]
110. Valtorta NK, Kanaan M, Gilbody S, Hanratty B. Loneliness, social isolation and social relationships: what are we measuring? A novel framework for classifying and comparing tools. *BMJ Open*. Apr 18, 2016;6(4):e010799. [doi: [10.1136/bmjopen-2015-010799](https://doi.org/10.1136/bmjopen-2015-010799)] [Medline: [27091822](https://pubmed.ncbi.nlm.nih.gov/27091822/)]
111. Simera I, Moher D, Hirst A, Hoey J, Schulz KF, Altman DG. Transparent and accurate reporting increases reliability, utility, and impact of your research: reporting guidelines and the EQUATOR Network. *BMC Med*. Apr 26, 2010;8(1):24. [doi: [10.1186/1741-7015-8-24](https://doi.org/10.1186/1741-7015-8-24)] [Medline: [20420659](https://pubmed.ncbi.nlm.nih.gov/20420659/)]
112. Choi M, Kong S, Jung D. Computer and internet interventions for loneliness and depression in older adults: a meta-analysis. *Healthc Inform Res*. Sep 2012;18(3):191-198. [doi: [10.4258/hir.2012.18.3.191](https://doi.org/10.4258/hir.2012.18.3.191)] [Medline: [23115742](https://pubmed.ncbi.nlm.nih.gov/23115742/)]
113. Chen YRR, Schulz PJ. The effect of information communication technology interventions on reducing social isolation in the elderly: a systematic review. *J Med Internet Res*. Jan 28, 2016;18(1):e18. [doi: [10.2196/jmir.4596](https://doi.org/10.2196/jmir.4596)] [Medline: [26822073](https://pubmed.ncbi.nlm.nih.gov/26822073/)]

114. Yen HY, Huang CW, Chiu HL, Jin G. The effect of social robots on depression and loneliness for older residents in long-term care facilities: a meta-analysis of randomized controlled trials. *J Am Med Dir Assoc*. Jun 2024;25(6):104979. [doi: [10.1016/j.jamda.2024.02.017](https://doi.org/10.1016/j.jamda.2024.02.017)] [Medline: [38614134](https://pubmed.ncbi.nlm.nih.gov/38614134/)]
115. Nowland R, Necka EA, Cacioppo JT. Loneliness and social internet use: pathways to reconnection in a digital world? *Perspect Psychol Sci*. Jan 2018;13(1):70-87. [doi: [10.1177/1745691617713052](https://doi.org/10.1177/1745691617713052)] [Medline: [28937910](https://pubmed.ncbi.nlm.nih.gov/28937910/)]
116. Lukas-novak/the-effectiveness-of-technology-based-interventions-on-the-reduction-of-loneliness-meta-analysis: fix of paths. Zenodo. URL: <https://zenodo.org/records/18061382> [Accessed 2026-04-16]

Abbreviations

CBT: cognitive behavioral therapy

GRADE: Grading of Recommendations Assessment, Development, and Evaluation

PRISMA: Preferred Reporting Items for Systematic Reviews and Meta-Analyses

PRISMA-S: Preferred Reporting Items for Systematic Reviews and Meta-Analyses-Search extension

PROSPERO: International Prospective Register of Systematic Reviews

RCT: randomized controlled trial

SMD: standardized mean difference

Edited by Stefano Brini; peer-reviewed by Hendrik Napierala, Zhen Liu; submitted 03.Jul.2025; final revised version received 03.Feb.2026; accepted 13.Feb.2026; published 08.May.2026

Please cite as:

Meier Z, Buchtova M, Sandora J, Novak L, Helvich J, Buchta O, Furstova J, Malinakova K, Tavel P

Efficacy of Technology-Based Interventions on the Reduction of Loneliness: Systematic Review and Meta-Analysis

J Med Internet Res 2026;28:e80059

URL: <https://www.jmir.org/2026/1/e80059>

doi: [10.2196/80059](https://doi.org/10.2196/80059)

© Zdenek Meier, Marie Buchtova, Jan Sandora, Lukas Novak, Jakub Helvich, Ondrej Buchta, Jana Furstova, Klara Malinakova, Peter Tavel. Originally published in the *Journal of Medical Internet Research* (<https://www.jmir.org>), 08.May.2026. This is an open-access article distributed under the terms of the Creative Commons Attribution License (<https://creativecommons.org/licenses/by/4.0/>), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work, first published in the *Journal of Medical Internet Research* (ISSN 1438-8871), is properly cited. The complete bibliographic information, a link to the original publication on <https://www.jmir.org/>, as well as this copyright and license information must be included.