

Review

Physical Activity Interventions Using Digital Health Interventions for Cancer-Related Fatigue in People With a History of Cancer: Scoping Review

Yeeun Kim^{1,2}, MSN; Ka Ryeong Bae³, PhD; Ji Hyun Sung⁴, PhD; Yun Hee Ko¹, PhD; Sue Kim⁵, PhD

¹College of Nursing, Yonsei University, Seoul, Republic of Korea

²School of Nursing, University of Texas at Austin, Austin, TX, United States

³College of Nursing, Eulji University, Seongnam, Republic of Korea

⁴College of Nursing, Keimyung University, Daegu, Republic of Korea

⁵Mo-Im Kim Nursing Research Institute, College of Nursing, Yonsei University, Seoul, Republic of Korea

Corresponding Author:

Sue Kim, PhD

Mo-Im Kim Nursing Research Institute

College of Nursing, Yonsei University

50-1, Yonsei-Ro, Seodaemun-gu

Seoul 03722

Republic of Korea

Phone: 82 2-2228-3276

Email: suekim@yuhs.ac

Abstract

Background: Although exercise has been proven effective in alleviating cancer-related fatigue (CRF), traditional face-to-face programs may not be accessible due to physical, temporal, or geographical barriers. Digital health interventions (DHIs) offer scalable alternatives for promoting physical activity; however, evidence synthesizing DHI-based physical activity interventions specifically targeting CRF and their intervention characteristics remains limited.

Objective: This scoping review aimed to map the types of digital health-based physical activity interventions for managing CRF, to summarize the key characteristics of DHI modalities and CRF outcomes, and to identify knowledge gaps for future research.

Methods: A systematic literature search was conducted across 4 databases (PubMed, EMBASE, Cochrane, and PsycINFO) up to December 2025. Inclusion criteria comprised experimental studies involving adults with a history of cancer, digital exercise interventions, a control group, and fatigue outcomes. Screening and data extraction followed the Joanna Briggs Institute Manual for Evidence Synthesis and PRISMA-ScR (Preferred Reporting Items for Systematic Reviews and Meta-Analyses Extension for Scoping Review) standards. The protocol was registered in PROSPERO (International Prospective Register of Systematic Reviews; CRD42022304285). Interventions were classified using the Evidence Standards Framework (ESF).

Results: Thirty-three studies comprising 3443 participants were included, representing 32 interventions. Most were randomized controlled trials (n=30, 91%). Interventions were delivered primarily via mobile apps (11/32, 34%) and wearable devices (8/32, 25%), followed by web-based platforms, videoconferencing, exergaming, and augmented reality. Eighteen interventions (reported in 19 studies) demonstrated statistically significant between-group CRF reductions, predominantly at immediate postintervention assessments. Evidence for sustained benefit beyond 12 months was limited, and only one study evaluated ultra-long-term outcomes, which did not demonstrate maintained improvement. Populations with breast cancer accounted for the largest proportion of participants. Fatigue measurement tools varied substantially, potentially contributing to heterogeneity in effect estimates. Most interventions were classified as ESF tier C, indicating a predominant focus on clinical outcome improvement rather than individual-level self-management and system-level implementation. Wearable device-based interventions showed the highest proportion of significant CRF improvement, whereas mobile app-based interventions reported both significant and nonsignificant results. DHIs lasting more than 12 weeks appeared to be associated with more effective CRF outcomes compared to shorter programs. Direct comparisons between in-person and digital delivery were scarce.

Conclusions:

Digital health-based physical activity interventions can reduce CRF in people with a history of cancer, with wearable device and longer-duration interventions showing the most favorable outcomes. However, the current evidence is concentrated in populations with breast cancer, and system-level integration remains unexplored. Future research should prioritize diverse populations with cancer, longer follow-up periods, standardized fatigue measurement, and systematic examination of specific intervention components that contribute to CRF reduction. Ultimately, advancing from individual efficacy trials toward scalable, workflow-integrated digital solutions will be key to sustainable CRF management across diverse oncology settings.

J Med Internet Res 2026;28:e83727; doi: [10.2196/83727](https://doi.org/10.2196/83727)

Keywords: people with a history of cancer; digital health; exercise; fatigue; neoplasm

Introduction

Cancer-related fatigue (CRF) is defined as “a distressing, persistent, subjective sense of physical, emotional, and/or cognitive tiredness or exhaustion related to cancer or cancer treatments that interfere with usual functioning” [1]. CRF is one of the most common symptoms of cancer and its treatment and persists for several years after cancer treatment. The prevalence of overall CRF was 43% among people with a history of cancer, and the prevalence of mild CRF was 71% [2]. Since one-third of people with a history of cancer have experienced affective, cognitive, or physical fatigue, CRF is associated with comorbidities, behavioral and psychological factors, and should be continuously assessed and addressed throughout the entire course of survivor care [3]. CRF has been associated with pain, depression, education level, and physical activity [3,4]. Accordingly, clinical practice guidelines, including those from the National Comprehensive Cancer Network, American Society of Clinical Oncology–Society for Integrative Oncology, and European Society for Medical Oncology, consistently recommend that CRF be routinely screened and that nonpharmacological interventions, particularly exercise, be offered as management strategies [1, 5,6].

Exercise is strongly recommended to alleviate CRF as a premier nonpharmacological intervention [5,7,8]. Multiple meta-analyses consistently suggested that exercise interventions significantly reduce CRF and improve the quality of life across various cancer types and treatment phases [7-9]. However, face-to-face exercise programs often encounter barriers related to accessibility, costs, and physical limitations of survivors [10,11]. These barriers have been further amplified for people with a history of cancer living in rural or underserved areas, and the COVID-19 pandemic has further highlighted the need for remote and flexible exercise delivery models [12]. This underscores the importance of exploring ways to deliver exercise interventions more feasibly and consistently throughout the survivorship period.

Digital health interventions (DHIs), defined as interventions delivered through digital technologies, have emerged as a scalable and accessible alternative, overcoming spatial and temporal constraints in cancer care [13,14]. In cancer care, DHIs have been applied across multiple domains, including symptom-focused interventions [15], promotion of physical activity through mobile apps [16], and delivery of psychological support via mobile health platforms [17]. These DHIs have been delivered through various modalities,

such as mobile apps, web pages, and wearable devices. These approaches may be particularly useful for extending supportive care beyond conventional in-person settings and supporting survivorship care. Given the established role of exercise in managing CRF, DHIs designed to promote physical activity are relevant for people with a history of cancer. However, it remains understudied which types of digital physical activity interventions are associated with improvements in CRF.

Several existing reviews have examined related topics, but each has notable limitations. Previous systematic reviews and meta-analyses of e-health interventions have focused on physical activity as the primary outcome without examining its effects on CRF [18,19]. Other reviews have been limited to specific cancer types, such as breast cancer, restricting the generalizability of their findings [19-21]. A recent scoping review indicated that the effectiveness of DHIs on fatigue varies considerably depending on the intervention’s type and functionality [22]. However, this review did not provide a granular synthesis of physical activity-specific digital interventions or clarify which components are most consistently associated with the reduction of CRF. To address these gaps, this scoping review specifically focuses on physical activity using DHIs, includes all cancer types, and maps the intervention components in relation to CRF. In addition, given the high degree of heterogeneity in digital platforms, exercise protocols, and fatigue assessment tools, a scoping review is the appropriate methodology for this emerging field [23]. It allows for a comprehensive mapping of the evidence and identifies specific intervention characteristics requiring further rigorous evaluation.

Despite the growing body of evidence supporting DHIs for promoting physical activity in people with a history of cancer, it remains understudied how the specific digital components of physical activity interventions relate to CRF across the diverse cancer population. Therefore, this scoping review aims to map the landscape of DHI-delivered physical activity interventions for managing CRF in people with a history of cancer, to summarize the key characteristics of the DHI modalities, CRF assessment methods, and reported CRF outcomes, and to identify knowledge gaps for future research.

Methods

Protocol and Registration

The protocol for this review was registered on PROSPERO (International Prospective Register of Systematic Reviews; registered number CRD42022304285). Although the protocol was initially planned as a systematic review, the study was conducted as a scoping review due to the high degree of heterogeneity in digital platforms, exercise protocols, and fatigue assessment tools in the included studies, which made a scoping review more appropriate for comprehensively mapping the evidence of physical activity DHI types for CRF.

Study Design

This scoping review selected studies on people with cancer engaging in physical activity DHIs targeting CRF and

Textbox 1. Population, concept, context

- Patient: People with a history of cancer (adults aged 18 years and older since diagnosis with cancer, excluding pediatric patients)
- Concept: exercise/physical activity through a digital health intervention (eg, mobile app, web-based, wearable device, etc) and cancer-related fatigue
- Context: open (no limitation)

The specific inclusion criteria were (1) experimental design studies with a control group, (2) using digital delivery methods, (3) employing a physical activity intervention for patients with a history of cancer, (4) reporting fatigue as an outcome, (5) aged 18 years and older at the time of diagnosis of cancer, and (6) studies published in English.

We excluded studies that did not meet one or more of these criteria, such as studies without a physical activity component, lacking digital delivery, not involving people with a history of cancer, without a fatigue outcome, or that were review or protocol papers. Multimodal interventions that included physical activity were included if fatigue outcomes linked to exercise were presented, whereas they were excluded if it was not possible to determine which intervention component influenced the outcome. Studies that used wearable devices solely for self-monitoring purposes and/or used phone follow-ups without interaction focused on the intervention were also excluded.

Information Sources

A systematic review of the major databases, such as PubMed, EMBASE, PsycINFO, and Cochrane, was primarily conducted until June 2024 and updated on January 6, 2026, for the period up to December 2025. References from relevant review papers were screened to identify additional eligible studies. The study registries, online resources, contacting authors, and additional information sources were not searched in line with the research purpose.

Search

The search strategy was reported following the PRISMA-Search (Preferred Reporting Items for Systematic Reviews and Meta-Analyses) [26] (Checklist 2). The search formulae

analyzed the findings. The study adhered to the guidelines of the JBI (Joanna Briggs Institute) Manual for Evidence Synthesis [24] and was reported in accordance with PRISMA-ScR (Preferred Reporting Items for Systematic Reviews and Meta-Analyses extension for Scoping Reviews) guidelines [25] (Checklist 1).

Eligibility Criteria

The key questions for the study were selected according to the JBI Manual for Evidence Synthesis [24] for population, concept, and context (Textbox 1).

comprised MeSH (Medical Subject Headings) terms and text words, with filtering methods employed to increase specificity based on the characteristics of each database. The search period was not limited for the review, which was based on key questions, and inclusion and exclusion criteria were used to select eligible studies. The search and data collection process was initially consulted with a professional librarian and reviewed and discussed in research team meetings. The research team comprised 2 PhDs, 1 graduate student, and 2 doctorally prepared professors of nursing. The existing search strategy was not reused due to inconsistent digital health keywords. After team discussions, a study-specific strategy was developed. The search terms combined the keywords “cancer,” “digital health,” and “cancer-related fatigue,” and the filters were for English and/or adults in each database (see Multimedia Appendix 1 for a full list of search terms). Two research team members (authors YK and YHK) independently conducted at least 1 search in each database individually to check for literature matches, and the retrieved papers were collected and organized using the EndNote 21 program.

Selection of Sources of Evidence

The screening of the collected data was conducted using Covidence (Veritas Health Innovation) [27] through more than 10 research team meetings. Duplicate records were removed using EndNote 21 (Clarivate) and Covidence. The titles and abstracts of the initially retrieved papers were reviewed independently, and then the full text of papers that appeared relevant was assessed. Inclusion and exclusion criteria were used to select eligible studies, but in cases of disagreement, the full text was reviewed with the entire team until consensus was reached through comprehensive discussion.

Data Charting Process

Data charting was conducted using Covidence, with 2 reviewers independently extracting data based on a predefined charting form developed from the study objectives.

The data were organized and categorized into tables according to the analysis framework that was created based on the objectives of the study. Studies were identified, categorized, and analyzed according to general characteristics (author, publication year, country, sample size, cancer type and stage, fatigue scale used, and intervention effectiveness) and intervention characteristics (intervention method and intervention delivery format). The Evidence Standards Framework (ESF) [28] was used to classify the DHI into 3 tiers: tier A targets system services to save health care providers' time and cost; tier B facilitates health management through communication and information, encompassing functional categories such as health care diaries and promoting good health; tier C involves clinical interventions through informing clinical management, driving clinical management, active monitoring, calculating, diagnosing, and treating conditions. In this study, we categorized as grade A if study outcomes included cost analysis of providers' time or expenses. If the intervention focused on offering general health information without interaction with experts, it was determined as grade B, and if it focused on personalized information or guidance for self-management or utilized interaction with experts to promote good health and healthy lifestyles, it was categorized as grade C.

Synthesis of Results

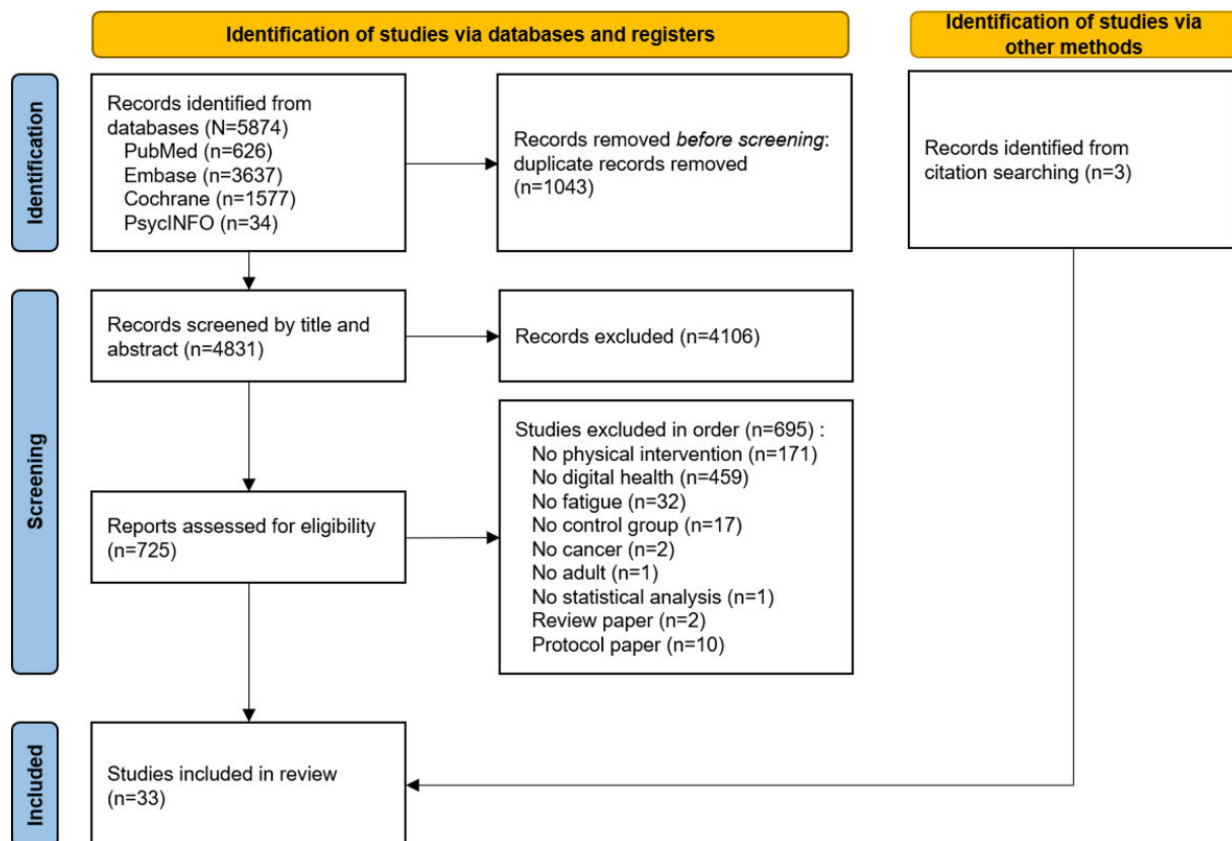
The results were synthesized descriptively to map the range and characteristics of digital health-based physical activity interventions targeting CRF. Studies were grouped and summarized according to study characteristics and intervention features, and frequencies were calculated for delivery modalities, intervention duration, ESF tiers, fatigue measurement tools, and intervention effectiveness.

Results

Selection of Sources of Evidence

A total of 5874 studies were identified based on the search strategy, and 3 studies were identified from citation searching. From this total, 1043 studies were removed due to duplications. Titles and abstracts of the remaining 4831 studies were screened based on the inclusion criteria. Consequently, 725 studies were excluded in the following order: no physical intervention (or unable to determine the effect of physical activity components; n=171), no digital health component in the intervention (n=459), no fatigue results (n=32), no control group (n=17), wrong population (n=2), no statistical analysis (n=1), review paper (n=2), and protocol paper (n=10). Ultimately, 33 studies [29-61] were included in the review (Figure 1).

Figure 1. PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analyses) 2020 flow diagram of study selection.



Characteristics of Sources of Evidence

The general characteristics of the 33 selected studies are shown in [Table 1](#) and [Multimedia Appendix 2](#). The publication dates ranged from 2015 to 2025, with 26 (79%) studies published after 2020. The United States accounted for the largest number of papers (n=9, 27%), followed, in order, by

China (n=5, 15%), Spain (n=4, 12%), Australia (n=3, 9%), the Netherlands (n=3, 9%), and other countries (n=1, 3% each). In terms of study design, the majority (n=30, 91%) were randomized controlled trials (RCTs), and only 3 (9%) were quasi-experimental designs. Most were 2-arm studies (n=29, 88%).

Table 1. Publication date, country, study design, and sample characteristics of the 33 selected studies.

Parameter and category	Values, n (%)	References
Publication date		
2019	7 (21)	[29-35]
2020-2025	26 (79)	[36-61]
Country		
USA	9 (27)	[37,39,41,45,51,54,56,57,59]
China	5 (15)	[34,48,50,55,61]
Spain	4 (12)	[30,42,46,53]
Australia	3 (9)	[36,49,58]
Netherlands	3 (9)	[32,38,44]
Brazil	2 (6)	[33,43]
Belgium	1 (3)	[43]
Canada	1 (3)	[29]
Denmark	1 (3)	[35]
Japan	1 (3)	[40]
Poland	1 (3)	[47]
South Korea	1 (3)	[31]
Taiwan	1 (3)	[60]
Study design		
RCT ^a	30 (91)	[29,30,32,34-36,38-61]
Non-RCT	3 (9)	[31,33,37]
Cancer type		
Single cancer		
Breast cancer	10 (30)	[30,31,36,40,42,46,50,51,53,60]
Hematologic cancer	3 (9)	[45,54,58]
Lung cancer	2 (6)	[52,55]
Others	6 (18)	[34,35,37,47,48,61]
Mixed cancers	5 (15)	[29,32,38,44,49]
All cancer types	7 (21)	[33,39,41,43,56,57,59]
Completion of cancer treatment		
Yes (survivor)	16 (48.5)	[29-31,36,38-40,42,45,46,49,52,53,57-59]
No (patient)	14 (42.4)	[33-35,37,41,43,47,48,50,51,54-56,61]
Mixed sample	3 (9.1)	[32,44,60]
Control group		
Usual care	18 (55)	[29-31,34,35,38,41,42,47,49,50,53,55-60]
Active care	7 (21)	[37,39,40,46,51,52,61]
Waitlist	5 (15)	[32,36,44,45,48]
Others	3 (9)	[33,43,54]

^aRCT: randomized controlled trial.

Upon examining key methodological characteristics of the included RCTs, most studies reported randomization and clearly defined control groups, whereas blinding was infrequently reported due to the behavioral nature of the interventions. Sample size and attrition were generally

reported, and validated fatigue measures and appropriate statistical analyses were consistently used.

All studies reported aiming to determine the feasibility and suitability of the developed interventions or to evaluate

their effectiveness. Two studies [42,44] were conducted to ascertain the long-term effects of the previous studies [30,32].

The total number of participants was 3443, with a range of 20 to 478 participants per study. When classified by cancer types, 21 out of 33 (63%) studies focused on a single type of cancer, of which the most common was breast cancer (10 studies). Five out of 33 (15%) studies included mixed samples of breast, prostate, and/or colorectal cancer, while 7 out of 33 (21%) were nonspecific and recruited all cancers.

Among the 33 studies included, 14 (42%) focused on patients receiving active cancer treatment, whereas 16 (48%) targeted people with a history of cancer who had completed intensive therapy, with less than half noting time since post treatment, which ranged from 1 month [53] to 2 years [45]. Three (9%) studies included participants in both treatment and posttreatment phases.

As presented in Table 1, for the control group, 18 (55%) studies used usual care and 5 (15%) employed a waitlist. Seven (21%) studies employed active care, such as providing control participants with wearable devices without further specification (n=3), allowing access to the mobile app (n=2),

and offering a coaching or education program (n=2). The remaining 3 studies were either a noncancer control [33], employed a crossover approach [43], or were factorial designs [54].

Results of Individual Sources of Evidence

Thirty-two interventions were analyzed, excluding studies [42,44] that reported on the same intervention [30,32]. As presented in Table 2 and Multimedia Appendix 3, the most frequent digital therapeutics and delivery method was a mobile app (n=11, 34%), of which most (n=9, 82%) were conducted after 2020, followed by wearable devices (n=8, 25%), such as wristband-type activity monitor devices. Other intervention modalities included a web page (n=5, 16%), videoconferencing (n=3, 9%), Xbox (n=3, 9%), augmented reality (n=1, 3%), and software that could not be classified as strictly mobile app-based or web-based (n=1, 3%). As shown in Figure 2, the number of studies using digital physical activity interventions increased since 2020. Wearable device and videoconferencing interventions emerged after 2020, whereas mobile apps and web page interventions were relatively distributed across the study period.

Table 2. Intervention characteristics description of the 32 interventions.

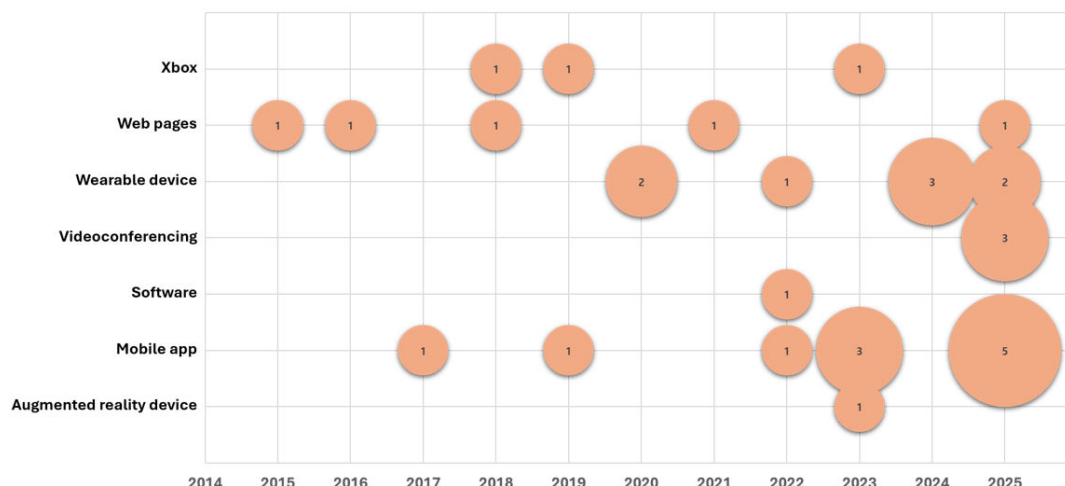
Parameter and category	Values, n (%)	References
Intervention modality		
Mobile app	11 (34)	[31,34,40,45,46,48,55-57,61]
Wearable device	8 (25)	[36,37,39,49-52,54]
Web pages	5 (16)	[29,30,32,38,59]
Videoconferencing	3 (9)	[53,58,60]
Xbox	3 (9)	[33,35,43]
Augmented reality device	1 (3)	[47]
Software	1 (3)	[41]
ESF^a grade		
B	2 (6)	[29,35]
C	30 (94)	[30-34,36-41,43,45-61]
Intervention duration		
<12 weeks	9 (28)	[29,30,33,43,45,46,52,58,61]
12 weeks	15 (47)	[31,35-37,39,40,48-51,56,57,59,60]
>12 weeks	8 (25)	[32,34,38,41,47,53-55]

^aESF: Evidence Standards Framework.

The intervention duration ranged from 3 weeks to 6 months, with 12 weeks (n=15, 47%) being the most frequently used period. The detailed intervention characteristics, including physical activity type and digital components, are presented in Multimedia Appendix 3. Most interventions combined digital activity tracking with some form of professional support, such as online coaching, while the specific combination of intervention components varied across studies.

Interventions were graded according to the ESF [28]. Two of the 32 (6%) interventions were grade B, that is, focused on supporting patients' self-management and wellness; the remaining were grade C, that is, treatment or diagnosis-focused interventions. There was no grade A, which aims to save cost and time for staff, which is likely the nature of the selected studies being focused on testing the efficacy of physical activity for people following cancer diagnosis.

Figure 2. Number of studies by year and intervention modality (n=32 interventions).



Various fatigue measurements were used in assessments (Table 3). The most frequently utilized scales were the Multidimensional Fatigue Inventory (MFI) and the European Organisation for Research and Treatment of Cancer Quality of Life Questionnaire C-30 (EORTC QLQ C-30) fatigue subscale, each used in 5 (15%) studies. Three (9%) studies each used one of the following: Functional Assessment of Chronic Illness Therapy-Fatigue Scale, Functional Assessment of Chronic Illness Therapy-Fatigue, Piper Fatigue Scale, Patient Reported Outcomes Measurement Information System, or the Functional Assessment of Cancer

Therapy-Fatigue or its fatigue subscale. Two (6%) studies each used either the Brief Fatigue Inventory, the Checklist Individual Strength, or the Chalder Fatigue Scale. Three studies used other instruments such as the Fatigue Symptom Inventory, the Multidimensional Fatigue Symptom Inventory—Short Form, and the Schwartz Cancer Fatigue Scale-6. All studies employed fatigue as the primary variable, with the exception of 4 studies [31,49,53,60], where the primary variable was quality of life, and fatigue was reported as a subscale of EORTC QLQ C-30.

Table 3. Description of fatigue measurements and statistical effects on fatigue between groups.

Parameter and category	Values, n (%)	References
Fatigue scale ^a		
MFI ^b	5 (15)	[37,38,48,52,61]
EORTC QLQ C-30 ^c	5 (15)	[31,49,53,55,60]
FACIT-Fatigue ^d	3 (9)	[36,45,47]
FACIT-F ^e	3 (9)	[33,43,58]
FACT-F ^f	3 (9)	[29,35,54]
PFS ^g	3 (9)	[30,42,46]
PROMIS ^h	3 (9)	[51,56,57]
BFI ⁱ	2 (6)	[34,59]
CIS ^j	2 (6)	[32,44]
CSF ^k	2 (6)	[40,50]
Fatigue Symptom Inventory	1 (3)	[39]
MSFI-SF ^l	1 (3)	[55]
SCFS-6 ^m	1 (3)	[41]
Statistically significant		
Yes (effective)	19 (58)	[30,32-34,36,37,39-41,43,44,46,50-55,60]
No (nonsignificant)	14 (42)	[29,31,35,38,42,45,47-49,56-59,61]

^aAs some studies used multiple scales, percentages sum to >100%.

^bMFI: Multidimensional Fatigue Inventory.

^cEORTC QLQ C-30: European Organisation for Research and Treatment of Cancer Quality of Life Questionnaire C-30.

^dFACIT-Fatigue: Functional Assessment of Chronic Illness Therapy-Fatigue Scale.

^eFACIT-F: Functional Assessment of Chronic Illness Therapy-Fatigue.

^fFACT-F: Functional Assessment of Cancer Therapy-Fatigue.

^gPFS: Piper Fatigue Scale.

^hPROMIS: Patient Reported Outcomes Measurement Information System.

^bBFI: Brief Fatigue Inventory.

^jCIS: Checklist Individual Strength.

^kCFS: Chalder Fatigue Scale.

^lMSFI-SF: Multidimensional Fatigue Symptom Inventory—Short Form.

^mSCFS-6: Schwartz Cancer Fatigue Scale-6.

In all studies, fatigue was measured a total of 44 times after baseline. Twenty-nine measurements occurred immediately at the conclusion of the intervention [29-31,33-41,43,45-48]. Four studies included midpoint measurements during the intervention [32-34,47], while 4 studies did not assess fatigue during the immediate postintervention period [32,42,44,58]. Twelve measurements focused on longer-term effects within 12 months after the postintervention period [30,32,36,44,46,49,58,59]. Another study evaluated an ultra-long-term effect 5 years later [42] from the original study [30].

Efficacy on Fatigue

As presented in Table 3, a total of 19 studies (18 interventions) reported a statistically significant reduction in CRF in the intervention group compared with the control group. Two studies employing the same intervention presented a positive effect on CRF 3 months before the end of the intervention and at 6 months [32,44] and 12 months after the intervention [44]. Building on the original report that showed significant results at the 6-month mark [30], its follow-up study [42]

examined outcomes 5 years after the intervention. However, no statistically significant improvement in fatigue was noted at this ultra-long-term assessment. Of the 18 interventions that reduced CRF, 7 (39%) were wearable devices, 4 (22%) were mobile apps, 2 (11%) each employed videoconferencing, web pages, or Xbox, and 1 (6%) used software as the DHI component.

As illustrated in the evidence gap map (Table 4), wearable device interventions showed the highest proportion of statistically significant CRF improvement (7/8, 88% interventions), whereas mobile app interventions showed the lowest proportion of that (4/10, 40%). No studies examined web page interventions for hematologic or lung cancer, and videoconferencing interventions were tested only for individuals with breast cancer and hematologic cancer. As shown in Figure 3, interventions lasting more than 12 weeks presented the highest proportion of significant CRF improvements (6/8, 75%), followed by those lasting less than 12 weeks (5/9, 56%) and those lasting 12 weeks (7/15, 47%).

Table 4. Evidence gap map of intervention modality by cancer type and cancer related fatigue across 33 included studies^a.

Intervention modality	Breast cancer	Hematologic cancer	Lung cancer	Others	Mixed cancers	All cancer types
Mobile app	[31,40,46]	[45]	[55]	[34,48,61]	— ^b	[56,57]
Wearable	[36,50,51]	[54]	[52]	[37]	[49]	[39]
Web pages	[30/42 ^c]	—	—	—	[29, 32/44 ^c , 38]	[59]
Videoconferencing	[53,60]	[58]	—	—	—	—
Xbox	—	—	—	[35]	—	[33,43]
AR ^d device	—	—	—	[47]	—	—
Software	—	—	—	—	—	[41]

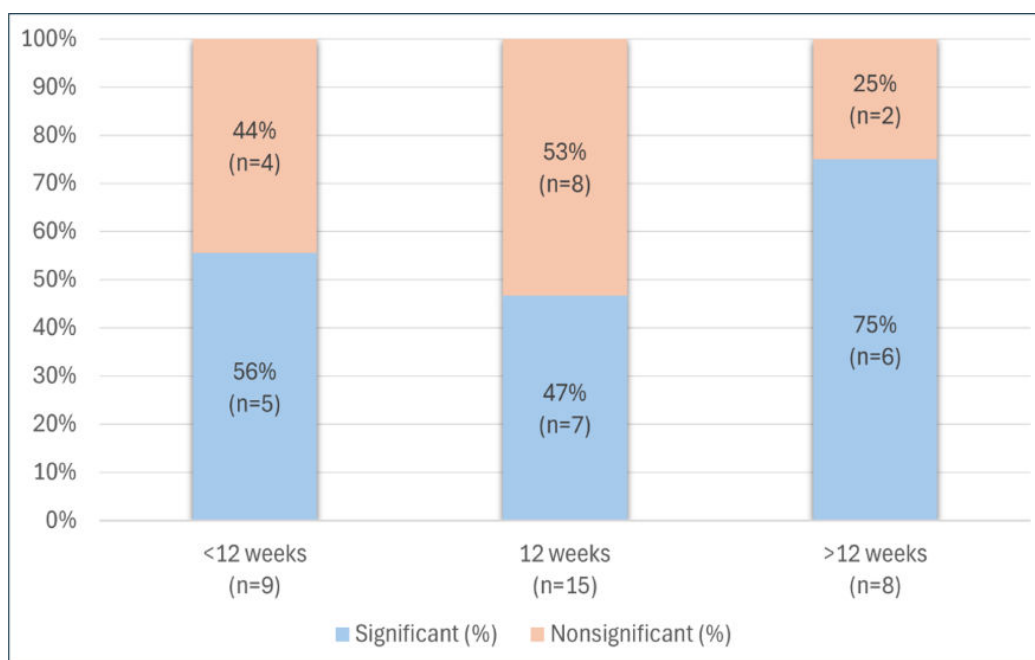
^aThe following studies showed statistically significant improvement in CRF between groups: [30,32-34,36,37,39-41,43,44,46,50-55,60].

^bNot available.

^cReferences [30] and [32] report the original intervention, and [42] and [44], respectively report its long-term follow-up.

^dAR: augmented reality.

Figure 3. Proportion of studies reporting statistically significant cancer-related fatigue (CRF) improvement by intervention duration (n=32 interventions).



Discussion

Principal Results

This scoping review mapped the current experimental evidence on physical activity programs delivered via DHIs to address CRF in people with a history of cancer and examined which modality of intervention features tend to improve in CRF. Mobile apps and wearable devices were the most commonly used delivery modalities, followed by web pages, videoconferencing, and exergaming. Twelve-week programs were the most frequently adopted intervention duration, and the evidence has expanded rapidly, with the majority of studies published after 2020, reflecting growing interest in digitally delivered exercise programs for people with a history of cancer. A wide range of fatigue assessment tools was used across the studies, with no single measurement predominating. More than half of the intervention reported statistically significant CRF improvements; however, findings at long-term follow-up were less consistent. Wearable device interventions showed the most consistent decrease in CRF, whereas mobile app interventions demonstrated more mixed findings. Interventions lasting more than 12 weeks were more effective in CRF than those of equal or less than 12 weeks.

Several knowledge gaps were identified. People with a history of breast cancer comprised the largest proportion of the study populations, and evidence for other cancer types, including hematologic or lung cancer, remained limited. Only one study assessed outcomes beyond 5 years, restricting the conclusion regarding the ultra-long-term durability of effects [42]. Furthermore, the high degree of heterogeneity in fatigue assessment tools, control group conditions, and digital delivery modalities made it difficult to directly compare intervention effects across the studies. These gaps highlighted the need for future studies examining diverse populations

with cancer, longer follow-up periods, and standardized fatigue assessments to better determine the optimal digital delivery modality and intervention components for CRF reduction.

Comparison With Prior Work

Previous systematic reviews and meta-analyses have primarily synthesized nondigital exercise interventions or relaxation-based approaches [7,8,62,63]. A meta-analysis [64] reported that eHealth interventions improved pain interference and sleep but not pain severity or fatigue among people with a history of cancer; however, exercise-specific digital interventions and sustainability were not systematically examined. Moreover, recent reviews [18,19,22] have typically aggregated digital interventions across heterogeneous behavioral targets (eg, symptom management, general physical activity, or psychosocial support), without distinguishing exercise-based programs or comparing platform-specific intervention characteristics.

The present review extends prior work by characterizing digital delivery modalities, mapping intervention scope using the ESF [28], and identifying heterogeneity across cancer types, treatment phases, and outcome measurements, thereby informing not only efficacy but also the implementation readiness of digital exercise interventions for CRF. ESF mapping shows that most interventions focus on individual-level support and pay relatively little attention to system-level efficiency and workflow integration. Mobile and wearable-based interventions primarily supported continuous self-monitoring and real-time feedback, which have been identified as core behavior change techniques facilitating awareness and engagement in physical activity promotion [13,14]. In contrast, web-based programs tended to rely more on structured educational content and scheduled interactions, which are typical characteristics of traditional web-based

health behavior interventions [65], potentially leading to different engagement patterns and intervention intensity. Most interventions were classified as ESF grade C, with no tier A interventions addressing system-level efficiency [28], indicating opportunities for scalable, implementation-oriented development, and integration into routine oncology service workflows. These modality-specific patterns suggest that the way support is delivered may influence participant engagement and the effective dose of the intervention.

These findings align with a recent meta-analysis, which reported that wearable electronic device system-supported physical activity programs significantly improved the quality of life and physical activity outcomes [66]. The continuous feedback mechanism inherent in wearable devices may enhance self-regulatory processes, thereby facilitating more consistent engagement with physical activity compared to other digital modalities. In contrast, although a previous review reported that mobile app-based interventions enhanced physical activity among people with a history of cancer [18], the present review found that mobile app interventions showed mixed results on CRF. This finding suggests that increased physical activity alone may not be sufficient to improve fatigue outcomes.

Fatigue trajectories after breast cancer treatment are influenced by biobehavioral factors, suggesting that intervention responsiveness may differ across diagnoses [67]. This concentration reflects where physical activity programs delivered via DHIs have been most feasible to test to date, underscoring the need to expand recruitment to more diverse populations with cancer. Given that treatment burden can influence baseline fatigue and responsiveness to exercise-based interventions [7,8,62], heterogeneity in cancer treatment exposure, including surgery, chemotherapy, radiotherapy, immunotherapy, and transplantation, may further constrain intervention outcomes—a consideration when analyzing physical activity DHIs. The mixed follow-up findings suggest that structured support initially improves outcomes, but maintenance may require continued reinforcement or transition-to-maintenance components after the program ends [54]. A recent review on digital health and telehealth in cancer care similarly reported that the evidence for fatigue outcomes was mixed [68]. This suggests that the inconsistent long-term results observed in the present review reflect a broader pattern in the field. Further research is needed to determine how long the CRF benefits of DHIs persist after the completion of intervention and what factors contribute to the maintenance or decrease of these effects over time. Direct comparisons between in-person and digital interventions remain limited, however, as only 2 of our selected studies explicitly compared delivery modalities [57,61]. One study comparing personalized exercise programs delivered in-person versus telehealth found that both interventions improved fatigue, without differences between interventions [57]. The other RCT reported improvement in physical activity with DHIs compared to in-person interventions, but no statistically significant fatigue differences between in-person and DHI modalities [61]. This suggests that whether interventions are delivered digitally

or in-person may not determine fatigue outcomes, but the intensity, duration, and content of interventions may be the crucial components of reducing fatigue mechanisms. This highlights a critical evidence gap regarding the comparative effectiveness and implementation efficiency of digital versus in-person interventions for CRF. Given the growing evidence supporting the association between physical activity and CRF in several people with a history of cancer, particularly people with a history of breast and prostate cancer [69], future study should examine whether DHIs on physical activity can produce consistent CRF benefits across more diverse populations with cancer and treatment settings.

Limitations

This review was limited to English-language publications. Methodological quality was examined using simplified descriptive indicators rather than formal risk-of-bias tools. Although multiple major biomedical databases were searched, the omission of engineering-focused databases (eg, IEEE Xplore) may have limited the identification of some technology-driven interventions.

The inclusion of diverse cancer diagnoses and heterogeneous treatment regimens, including differences in treatment intensity and recovery trajectories, may have influenced baseline fatigue severity and intervention responsiveness, limiting interpretability across populations. People with a history of breast cancer comprised the largest proportion of study populations, limiting generalizability to other cancer types and male populations. In addition, long-term follow-up evidence remains scarce, with only one study assessing outcomes beyond 5 years [42], restricting conclusions regarding the durability of effects. Finally, variability in digital access and literacy may further affect real-world applicability of these findings.

Heterogeneity across the included studies may have influenced the observed results in several ways. First, a wide range of fatigue assessment tools was used across studies, with no single instrument predominating, which limits the direct comparison of CRF outcomes across studies. Second, control conditions varied substantially, ranging from usual care and waitlist controls to active comparators using wearable devices or mobile apps, which may have attenuated detectable intervention effects and limited cross-study comparability. Third, the wide age range of participants, spanning young adults to older survivors, may further contribute to variability in engagement and intervention responsiveness due to differences in digital familiarity and physical capacity [70].

Conclusions

This scoping review identified several major types of physical activity DHIs for CRF in people with a history of cancer and found that wearable device interventions and longer-duration interventions represented an effective approach for CRF outcomes. Furthermore, as the demand for DHI modalities that actually prompt people with a history of cancer to move may increase in the future, our findings can be used by practitioners as they consider selecting and applying

digital health tools to deliver activity-based interventions for alleviating CRF. As the reviewed DHIs were predominantly on patients with breast cancer, clinicians and researchers can respond to the shortcomings identified in this review by actively exploring strategies to engage and include diverse cancer types and underserved populations, while including CRF as a primary outcome.

Overall, this review supports the potential clinical relevance of digital health-based exercise interventions for managing CRF. Beyond individual-level efficacy, the

absence of system-level integration (ESF grade A) across the reviewed interventions points to a critical gap between current research and real-world implementation. To enhance real-world impact, future research should prioritize standardized fatigue measurements, broader populations with cancer, and strategies to bridge the digital health literacy gap. Ultimately, shifting the focus from individual efficacy to scalable, workflow-integrated digital solutions will be essential for the sustainable management of CRF in diverse oncology settings.

Acknowledgments

We gratefully acknowledge Professor Sun Young Park (College of Nursing, Kyungpook National University, Korea) for creating the original search strategy and Dr. Na Won Kim and Ms. Dami Jeong (Yonsei University Medical Library) for reviewing the search process. We used the generative artificial intelligence tool ChatGPT by OpenAI under human supervision for 2 tasks classified according to the CAIDeT taxonomy: validation, which involved cross-checking the tables, and summarizing text, which involved assisting with the first draft of the abstract. Responsibility for the final manuscript lies entirely with the authors. Generative artificial intelligence tools are not listed as authors and do not bear responsibility for the final outcomes.

Funding

This work was supported by a 2021 Alumni Research Cluster Grant (number 6-2021-0181) from Yonsei University College of Nursing. No funder had any role in the study design or decision to submit the paper for publication.

Data Availability

All data generated within this scoping review are available in tables and attached as multimedia appendices, such as search strategy and details of the selected studies.

Authors' Contributions

Conceptualization: YK, SK (equal)

Data curation: YK

Formal analysis: YK (lead), YHK (equal), KRB (supporting), JHS (supporting), SK (supporting)

Funding acquisition: SK

Investigation: YK, KRB, JHS, YHK, SK (equal)

Methodology: YK, KRB, JHS, YHK, SK (equal)

Project administration: YK (lead), YHK (supporting)

Supervision: SK

Visualization: YK (lead), YHK (supporting)

Writing – original draft: KRB (lead), YK (supporting)

Writing – review and editing: SK (lead), JHS (supporting), YHK (supporting)

Conflicts of interest

None declared.

Multimedia Appendix 1

Full list of search terms.

[\[XLSX File \(Microsoft Excel File\), 20 KB-Multimedia Appendix 1\]](#)

Multimedia Appendix 2

General characteristics, study aims, and population characteristics of included studies.

[\[DOCX File \(Microsoft Word File\), 38 KB-Multimedia Appendix 2\]](#)

Multimedia Appendix 3

Summary of intervention characteristics of physical activity and digital health components.

[\[DOCX File \(Microsoft Word File\), 35 KB-Multimedia Appendix 3\]](#)

Checklist 1

PRISMA ScR checklist.

[\[DOCX File \(Microsoft Word File\), 53 KB-Checklist 1\]](#)

Checklist 2

PRISMA search checklist.

[\[DOCX File \(Microsoft Word File\), 19 KB-Checklist 2\]](#)**References**

1. Berger AM, Mooney K, Alvarez-Perez A, et al. Cancer-related fatigue, version 2.2015. *J Natl Compr Canc Netw*. Aug 2015;13(8):1012-1039. [doi: [10.6004/jnccn.2015.0122](https://doi.org/10.6004/jnccn.2015.0122)] [Medline: [26285247](https://pubmed.ncbi.nlm.nih.gov/26285247/)]
2. Kang YE, Yoon JH, Park NH, Ahn YC, Lee EJ, Son CG. Prevalence of cancer-related fatigue based on severity: a systematic review and meta-analysis. *Sci Rep*. Aug 7, 2023;13(1):12815. [doi: [10.1038/s41598-023-39046-0](https://doi.org/10.1038/s41598-023-39046-0)] [Medline: [37550326](https://pubmed.ncbi.nlm.nih.gov/37550326/)]
3. Thong MSY, Doege D, Koch-Gallenkamp L, et al. Fatigue in long-term cancer survivors: prevalence, associated factors, and mortality. A prospective population-based study. *Br J Cancer*. Oct 2025;133(6):831-843. [doi: [10.1038/s41416-025-03116-z](https://doi.org/10.1038/s41416-025-03116-z)] [Medline: [40665014](https://pubmed.ncbi.nlm.nih.gov/40665014/)]
4. Zeilinger EL, Znic-Novakovic I, Oppenauer C, et al. Prevalence and biopsychosocial indicators of fatigue in cancer patients. *Cancer Med*. Jun 2024;13(11):e7293. [doi: [10.1002/cam4.7293](https://doi.org/10.1002/cam4.7293)] [Medline: [38819432](https://pubmed.ncbi.nlm.nih.gov/38819432/)]
5. Bower JE, Lacchetti C, Alici Y, et al. Management of fatigue in adult survivors of cancer: ASCO-Society for Integrative Oncology guideline update. *J Clin Oncol*. Jul 10, 2024;42(20):2456-2487. [doi: [10.1200/JCO.24.00541](https://doi.org/10.1200/JCO.24.00541)] [Medline: [38754041](https://pubmed.ncbi.nlm.nih.gov/38754041/)]
6. Fabi A, Bhargava R, Fatigoni S, et al. Cancer-related fatigue: ESMO Clinical Practice Guidelines for diagnosis and treatment. *Ann Oncol*. Jun 2020;31(6):713-723. [doi: [10.1016/j.annonc.2020.02.016](https://doi.org/10.1016/j.annonc.2020.02.016)] [Medline: [32173483](https://pubmed.ncbi.nlm.nih.gov/32173483/)]
7. Chen X, Li J, Chen C, et al. Effects of exercise interventions on cancer-related fatigue and quality of life among cancer patients: a meta-analysis. *BMC Nurs*. Jun 13, 2023;22(1):200. [doi: [10.1186/s12912-023-01363-0](https://doi.org/10.1186/s12912-023-01363-0)] [Medline: [37312185](https://pubmed.ncbi.nlm.nih.gov/37312185/)]
8. Herranz-Gómez A, Cuenca-Martínez F, Suso-Martí L, et al. Effectiveness of therapeutic exercise models on cancer-related fatigue in patients with cancer undergoing chemotherapy: a systematic review and network meta-analysis. *Arch Phys Med Rehabil*. Aug 2023;104(8):1331-1342. [doi: [10.1016/j.apmr.2023.01.008](https://doi.org/10.1016/j.apmr.2023.01.008)] [Medline: [36736602](https://pubmed.ncbi.nlm.nih.gov/36736602/)]
9. Zhou S, Chen G, Xu X, et al. Comparative efficacy of various exercise types on cancer-related fatigue for cancer survivors: a systematic review and network meta-analysis of randomized controlled trials. *Cancer Med*. Apr 2025;14(7):e70816. [doi: [10.1002/cam4.70816](https://doi.org/10.1002/cam4.70816)] [Medline: [40145635](https://pubmed.ncbi.nlm.nih.gov/40145635/)]
10. Soriano-Maldonado A, Díez-Fernández DM, Esteban-Simón A, et al. Effects of a 12-week supervised resistance training program, combined with home-based physical activity, on physical fitness and quality of life in female breast cancer survivors: the EFICAN randomized controlled trial. *J Cancer Surviv*. Oct 2023;17(5):1371-1385. [doi: [10.1007/s11764-022-01192-1](https://doi.org/10.1007/s11764-022-01192-1)] [Medline: [35314958](https://pubmed.ncbi.nlm.nih.gov/35314958/)]
11. Lin Y, Wu C, He C, et al. Effectiveness of three exercise programs and intensive follow-up in improving quality of life, pain, and lymphedema among breast cancer survivors: a randomized, controlled 6-month trial. *Support Care Cancer*. Dec 13, 2022;31(1):9. [doi: [10.1007/s00520-022-07494-5](https://doi.org/10.1007/s00520-022-07494-5)] [Medline: [36512157](https://pubmed.ncbi.nlm.nih.gov/36512157/)]
12. Lopez AM. Telehealth in cancer care: inequities, barriers, and opportunities. *Cancer J*. 2024;30(1):2-7. [doi: [10.1097/PPO.0000000000000694](https://doi.org/10.1097/PPO.0000000000000694)] [Medline: [38265919](https://pubmed.ncbi.nlm.nih.gov/38265919/)]
13. Jahnel T, Pan CC, Pedros Barnils N, et al. Developing and evaluating digital public health interventions using the digital public health framework DigiPHrame: a framework development study. *J Med Internet Res*. Sep 12, 2024;26:e54269. [doi: [10.2196/54269](https://doi.org/10.2196/54269)] [Medline: [39264696](https://pubmed.ncbi.nlm.nih.gov/39264696/)]
14. Kasoju N, Remya NS, Sasi R, et al. Digital health: trends, opportunities and challenges in medical devices, pharma and bio-technology. *CSI Trans ICT*. Apr 2023;11(1):11-30. [doi: [10.1007/s40012-023-00380-3](https://doi.org/10.1007/s40012-023-00380-3)]
15. Mayer C, Walch O, Dempsey W, et al. A circadian and app-based personalized lighting intervention for the reduction of cancer-related fatigue. *Cell Rep Med*. Mar 18, 2025;6(3):102001. [doi: [10.1016/j.xcrm.2025.102001](https://doi.org/10.1016/j.xcrm.2025.102001)] [Medline: [40056908](https://pubmed.ncbi.nlm.nih.gov/40056908/)]
16. Cheung DST, Kwok TWH, Liu S, et al. A smartphone app (WExercise) to promote physical activity among cancer survivors: randomized controlled trial. *J Med Internet Res*. Oct 3, 2025;27:e75839. [doi: [10.2196/75839](https://doi.org/10.2196/75839)] [Medline: [41042984](https://pubmed.ncbi.nlm.nih.gov/41042984/)]
17. Barth J, Schlöpfer S, Schneider F, et al. Mobile health intervention CanRelax reduces distress in people with cancer in a randomized controlled trial. *NPJ Digit Med*. May 10, 2025;8(1):269. [doi: [10.1038/s41746-025-01688-x](https://doi.org/10.1038/s41746-025-01688-x)] [Medline: [40348854](https://pubmed.ncbi.nlm.nih.gov/40348854/)]
18. Xiao K, Tang L, Chen Y, Zhou J, Yang Q, Wang R. The effectiveness of E-health interventions promoting physical activity in cancer survivors: a systematic review and meta-analysis of randomized controlled trials. *J Cancer Res Clin Oncol*. Feb 2, 2024;150(2):72. [doi: [10.1007/s00432-023-05546-9](https://doi.org/10.1007/s00432-023-05546-9)] [Medline: [38305910](https://pubmed.ncbi.nlm.nih.gov/38305910/)]

19. Kang H, Moon M. Effects of digital physical activity interventions for breast cancer patients and survivors: a systematic review and meta-analysis. *Healthc Inform Res*. Oct 2023;29(4):352-366. [doi: [10.4258/hir.2023.29.4.352](https://doi.org/10.4258/hir.2023.29.4.352)] [Medline: [37964457](https://pubmed.ncbi.nlm.nih.gov/37964457/)]
20. Cooper KB, Lapierre S, Carrera Seoane M, et al. Behavior change techniques in digital physical activity interventions for breast cancer survivors: a systematic review. *Transl Behav Med*. Apr 15, 2023;13(4):268-280. [doi: [10.1093/tbm/ibac111](https://doi.org/10.1093/tbm/ibac111)] [Medline: [36694356](https://pubmed.ncbi.nlm.nih.gov/36694356/)]
21. Zhang X, Fang J, Hao Y, Yang D, Luo J, Li X. Digital behavior change interventions to promote physical activity and reduce sedentary behavior among survivors of breast cancer: systematic review and meta-analysis of randomized controlled trials. *J Med Internet Res*. Jun 19, 2025;27:e65278. [doi: [10.2196/65278](https://doi.org/10.2196/65278)] [Medline: [40537089](https://pubmed.ncbi.nlm.nih.gov/40537089/)]
22. Lee K, Kim S, Kim SH, et al. Digital health interventions for adult patients with cancer evaluated in randomized controlled trials: scoping review. *J Med Internet Res*. Jan 6, 2023;25:e38333. [doi: [10.2196/38333](https://doi.org/10.2196/38333)] [Medline: [36607712](https://pubmed.ncbi.nlm.nih.gov/36607712/)]
23. Peters MDJ, Marnie C, Tricco AC, et al. Updated methodological guidance for the conduct of scoping reviews. *JB I Evid Synth*. Oct 2020;18(10):2119-2126. [doi: [10.11124/JBIES-20-00167](https://doi.org/10.11124/JBIES-20-00167)] [Medline: [33038124](https://pubmed.ncbi.nlm.nih.gov/33038124/)]
24. Aromataris E, Lockwood C, Porritt K, Pilla B, Jordan Z, editors. *JB I Manual for Evidence Synthesis*. Joanna Briggs Institute (JBI); 2024. [doi: [10.46658/JBIMES-24-01](https://doi.org/10.46658/JBIMES-24-01)]
25. Tricco AC, Lillie E, Zarin W, et al. PRISMA extension for Scoping Reviews (PRISMA-ScR): checklist and explanation. *Ann Intern Med*. Oct 2, 2018;169(7):467-473. [doi: [10.7326/M18-0850](https://doi.org/10.7326/M18-0850)] [Medline: [30178033](https://pubmed.ncbi.nlm.nih.gov/30178033/)]
26. Rethlefsen ML, Page MJ. PRISMA 2020 and PRISMA-S: common questions on tracking records and the flow diagram. *J Med Libr Assoc*. Apr 1, 2022;110(2):253-257. [doi: [10.5195/jmla.2022.1449](https://doi.org/10.5195/jmla.2022.1449)] [Medline: [35440907](https://pubmed.ncbi.nlm.nih.gov/35440907/)]
27. Covidence. URL: <https://www.covidence.org> [Accessed 2025-08-24]
28. Evidence standards framework for digital health technologies. National Institute for Health and Care Excellence (NICE); 2018. URL: <https://www.nice.org.uk/corporate/eecd7/resources/evidence-standards-framework-for-digital-health-technologies-pdf-1124017457605> [Accessed 2025-04-01]
29. Forbes CC, Blanchard CM, Mummery WK, Courneya KS. Feasibility and preliminary efficacy of an online intervention to increase physical activity in Nova Scotian cancer survivors: a randomized controlled trial. *JMIR Cancer*. Nov 23, 2015;1(2):e12. [doi: [10.2196/cancer.4586](https://doi.org/10.2196/cancer.4586)] [Medline: [28410166](https://pubmed.ncbi.nlm.nih.gov/28410166/)]
30. Galiano-Castillo N, Cantarero-Villanueva I, Fernández-Lao C, et al. Telehealth system: a randomized controlled trial evaluating the impact of an internet-based exercise intervention on quality of life, pain, muscle strength, and fatigue in breast cancer survivors. *Cancer*. Oct 15, 2016;122(20):3166-3174. [doi: [10.1002/ncr.30172](https://doi.org/10.1002/ncr.30172)] [Medline: [27332968](https://pubmed.ncbi.nlm.nih.gov/27332968/)]
31. Uhm KE, Yoo JS, Chung SH, et al. Effects of exercise intervention in breast cancer patients: is mobile health (mHealth) with pedometer more effective than conventional program using brochure? *Breast Cancer Res Treat*. Feb 2017;161(3):443-452. [doi: [10.1007/s10549-016-4065-8](https://doi.org/10.1007/s10549-016-4065-8)] [Medline: [27933450](https://pubmed.ncbi.nlm.nih.gov/27933450/)]
32. Golsteijn RHJ, Bolman C, Volders E, Peels DA, de Vries H, Lechner L. Short-term efficacy of a computer-tailored physical activity intervention for prostate and colorectal cancer patients and survivors: a randomized controlled trial. *Int J Behav Nutr Phys Act*. Oct 30, 2018;15(1):106. [doi: [10.1186/s12966-018-0734-9](https://doi.org/10.1186/s12966-018-0734-9)] [Medline: [30376857](https://pubmed.ncbi.nlm.nih.gov/30376857/)]
33. Oliveira PF, Iunes DH, Alves RS, Carvalho JM, Menezes FS, Carvalho LC. Effects of exergaming in cancer related fatigue in the quality of life and electromyography of the middle deltoid of people with cancer in treatment: a controlled trial. *Asian Pac J Cancer Prev*. Sep 26, 2018;19(9):2591-2597. [doi: [10.22034/APJCP.2018.19.9.2591](https://doi.org/10.22034/APJCP.2018.19.9.2591)] [Medline: [30256065](https://pubmed.ncbi.nlm.nih.gov/30256065/)]
34. Lu Y, Qu HQ, Chen FY, et al. Effect of Baduanjin qigong exercise on cancer-related fatigue in patients with colorectal cancer undergoing chemotherapy: a randomized controlled trial. *Oncol Res Treat*. 2019;42(9):431-439. [doi: [10.1159/000501127](https://doi.org/10.1159/000501127)] [Medline: [31266043](https://pubmed.ncbi.nlm.nih.gov/31266043/)]
35. Villumsen BR, Jorgensen MG, Frystyk J, Hørdam B, Borre M. Home-based “exergaming” was safe and significantly improved 6-min walking distance in patients with prostate cancer: a single-blinded randomised controlled trial. *BJU Int*. Oct 2019;124(4):600-608. [doi: [10.1111/bju.14782](https://doi.org/10.1111/bju.14782)] [Medline: [31012238](https://pubmed.ncbi.nlm.nih.gov/31012238/)]
36. Vallance JK, Nguyen NH, Moore MM, et al. Effects of the ACTIVity And TEchnology (ACTIVATE) intervention on health-related quality of life and fatigue outcomes in breast cancer survivors. *Psychooncology*. Jan 2020;29(1):204-211. [doi: [10.1002/pon.5298](https://doi.org/10.1002/pon.5298)] [Medline: [31763746](https://pubmed.ncbi.nlm.nih.gov/31763746/)]
37. Xiao C, Beitler JJ, Higgins KA, et al. Pilot study of combined aerobic and resistance exercise on fatigue for patients with head and neck cancer: inflammatory and epigenetic changes. *Brain Behav Immun*. Aug 2020;88:184-192. [doi: [10.1016/j.bbi.2020.04.044](https://doi.org/10.1016/j.bbi.2020.04.044)] [Medline: [32330594](https://pubmed.ncbi.nlm.nih.gov/32330594/)]
38. van de Wiel HJ, Stuver MM, May AM, et al. Effects of and lessons learned from an internet-based physical activity support program (with and without physiotherapist telephone counselling) on physical activity levels of breast and prostate cancer survivors: the PABLO randomized controlled trial. *Cancers (Basel)*. Jul 21, 2021;13(15):3665. [doi: [10.3390/cancers13153665](https://doi.org/10.3390/cancers13153665)] [Medline: [34359567](https://pubmed.ncbi.nlm.nih.gov/34359567/)]

39. Johnson AM, Baker KS, Haviland MJ, et al. A pilot randomized controlled trial of a Fitbit- and Facebook-based physical activity intervention for young adult cancer survivors. *J Adolesc Young Adult Oncol*. Aug 2022;11(4):379-388. [doi: [10.1089/jayao.2021.0056](https://doi.org/10.1089/jayao.2021.0056)] [Medline: [34677081](https://pubmed.ncbi.nlm.nih.gov/34677081/)]
40. Ochi E, Tsuji K, Narisawa T, et al. Cardiorespiratory fitness in breast cancer survivors: a randomised controlled trial of home-based smartphone supported high intensity interval training. *BMJ Support Palliat Care*. Mar 2022;12(1):33-37. [doi: [10.1136/bmjspcare-2021-003141](https://doi.org/10.1136/bmjspcare-2021-003141)] [Medline: [34389552](https://pubmed.ncbi.nlm.nih.gov/34389552/)]
41. Wilkie DJ, Schwartz AL, Liao WC, et al. Reduced cancer-related fatigue after tablet-based exercise education for patients. *Cancer Control*. 2022;29:10732748221087054. [doi: [10.1177/10732748221087054](https://doi.org/10.1177/10732748221087054)] [Medline: [35414203](https://pubmed.ncbi.nlm.nih.gov/35414203/)]
42. Álvarez-Salvago F, Jiménez-García JD, Martínez-Amat A, et al. Does participation in therapeutic exercise programs after finishing oncology treatment still ensure an adequate health status for long-term breast cancer survivors? A ≥ 5 years follow-up study. *Support Care Cancer*. May 18, 2023;31(6):343. [doi: [10.1007/s00520-023-07801-8](https://doi.org/10.1007/s00520-023-07801-8)] [Medline: [37199790](https://pubmed.ncbi.nlm.nih.gov/37199790/)]
43. da Silva Alves R, de Carvalho JM, Borges JBC, Nogueira DA, Iunes DH, Carvalho LC. Effect of exergaming on quality of life, fatigue, and strength and endurance muscle in cancer patients: a randomized crossover trial. *Games Health J*. Oct 2023;12(5):358-365. [doi: [10.1089/g4h.2022.0161](https://doi.org/10.1089/g4h.2022.0161)] [Medline: [37155685](https://pubmed.ncbi.nlm.nih.gov/37155685/)]
44. Golsteijn RHJ, Bolman C, Peels DA, Volders E, de Vries H, Lechner L. Long-term efficacy of a computer-tailored physical activity intervention for prostate and colorectal cancer patients and survivors: a randomized controlled trial. *J Sport Health Sci*. Nov 2023;12(6):690-704. [doi: [10.1016/j.jshs.2023.08.002](https://doi.org/10.1016/j.jshs.2023.08.002)] [Medline: [37591482](https://pubmed.ncbi.nlm.nih.gov/37591482/)]
45. Lee K, Shamunee J, Lindenfeld L, et al. Feasibility of implementing a supervised telehealth exercise intervention in frail survivors of hematopoietic cell transplantation: a pilot randomized trial. *BMC Cancer*. May 1, 2023;23(1):390. [doi: [10.1186/s12885-023-10884-5](https://doi.org/10.1186/s12885-023-10884-5)] [Medline: [37127595](https://pubmed.ncbi.nlm.nih.gov/37127595/)]
46. Lozano-Lozano M, Galiano-Castillo N, Gonzalez-Santos A, et al. Effect of mHealth plus occupational therapy on cognitive function, mood and physical function in people after cancer: secondary analysis of a randomized controlled trial. *Ann Phys Rehabil Med*. Mar 2023;66(2):101681. [doi: [10.1016/j.rehab.2022.101681](https://doi.org/10.1016/j.rehab.2022.101681)] [Medline: [35671976](https://pubmed.ncbi.nlm.nih.gov/35671976/)]
47. Pieczyńska A, Zasadzka E, Pilarska A, Procyk D, Adamska K, Hojan K. Rehabilitation exercises supported by monitor-augmented reality for patients with high-grade glioma undergoing radiotherapy: results of a randomized clinical trial. *J Clin Med*. Oct 29, 2023;12(21):6838. [doi: [10.3390/jcm12216838](https://doi.org/10.3390/jcm12216838)] [Medline: [37959303](https://pubmed.ncbi.nlm.nih.gov/37959303/)]
48. Wen L, Chen X, Cui Y, Zhang M, Bai X. Effects of Baduanjin exercise in nasopharyngeal carcinoma patients after chemoradiotherapy: a randomized controlled trial. *Support Care Cancer*. Dec 23, 2022;31(1):79. [doi: [10.1007/s00520-022-07548-8](https://doi.org/10.1007/s00520-022-07548-8)] [Medline: [36562869](https://pubmed.ncbi.nlm.nih.gov/36562869/)]
49. Hardcastle SJ, Leyton-Román M, Maxwell-Smith C, Hince D. Impact of the promoting physical activity in regional and remote cancer survivors intervention on health-related quality of life in breast and colorectal cancer survivors. *Front Oncol*. 2024;14:1368119. [doi: [10.3389/fonc.2024.1368119](https://doi.org/10.3389/fonc.2024.1368119)] [Medline: [39309736](https://pubmed.ncbi.nlm.nih.gov/39309736/)]
50. Li H, Sang D, Gong L, et al. Improving physical and mental health in women with breast cancer undergoing anthracycline-based chemotherapy through wearable device-based aerobic exercise: a randomized controlled trial. *Front Public Health*. 2024;12:1451101. [doi: [10.3389/fpubh.2024.1451101](https://doi.org/10.3389/fpubh.2024.1451101)] [Medline: [39363984](https://pubmed.ncbi.nlm.nih.gov/39363984/)]
51. Phillips SM, Starikovskiy J, Solk P, et al. Feasibility and preliminary effects of the Fit2ThriveMB pilot physical activity promotion intervention on physical activity and patient reported outcomes in individuals with metastatic breast cancer. *Breast Cancer Res Treat*. Nov 2024;208(2):391-403. [doi: [10.1007/s10549-024-07432-5](https://doi.org/10.1007/s10549-024-07432-5)] [Medline: [39014267](https://pubmed.ncbi.nlm.nih.gov/39014267/)]
52. Arents E, Haesevoets S, Hermans F, et al. Physical activity telecoaching in post-surgical NSCLC patients: a mixed-methods pilot study exploring feasibility, acceptability and actual usage. *Cancers (Basel)*. Sep 2, 2025;17(17):2886. [doi: [10.3390/cancers17172886](https://doi.org/10.3390/cancers17172886)] [Medline: [40940983](https://pubmed.ncbi.nlm.nih.gov/40940983/)]
53. Lavín-Pérez AM, Collado-Mateo D, Nieto I, et al. Effects of individualized high-intensity online concurrent exercise guided by autonomic modulation on the mental health and quality of life of breast cancer survivors. *Psychooncology*. Dec 2025;34(12):e70348. [doi: [10.1002/pon.70348](https://doi.org/10.1002/pon.70348)] [Medline: [41339114](https://pubmed.ncbi.nlm.nih.gov/41339114/)]
54. Lee CY, Gordon MJ, Markofski MM, et al. Optimization of mHealth behavioral interventions for patients with chronic lymphocytic leukemia: the HEALTH4CLL study. *J Cancer Surviv*. Aug 2025;19(4):1325-1334. [doi: [10.1007/s11764-024-01555-w](https://doi.org/10.1007/s11764-024-01555-w)] [Medline: [38472612](https://pubmed.ncbi.nlm.nih.gov/38472612/)]
55. Li G, Zhou X, Deng J, et al. Digital therapeutics-based cardio-oncology rehabilitation for lung cancer survivors: randomized controlled trial. *JMIR mHealth uHealth*. Feb 25, 2025;13:e60115. [doi: [10.2196/60115](https://doi.org/10.2196/60115)] [Medline: [39999435](https://pubmed.ncbi.nlm.nih.gov/39999435/)]
56. Lukkahatai N, Benjasirisan C, Shen A, et al. Combined technology-enhanced home exercise and acupuncture (TEHEplus) program on symptoms among cancer patients receiving immunotherapy: a feasibility study. *BMC Cancer*. Oct 1, 2025;25(1):1481. [doi: [10.1186/s12885-025-14887-2](https://doi.org/10.1186/s12885-025-14887-2)] [Medline: [41034793](https://pubmed.ncbi.nlm.nih.gov/41034793/)]

57. Lukkahatai N, Han G, Benjasirisan C, et al. A comparison of in-person and telehealth personalized exercise programs for cancer survivors: a secondary data analysis. *Cancers (Basel)*. Jul 23, 2025;17(15):2432. [doi: [10.3390/cancers17152432](https://doi.org/10.3390/cancers17152432)] [Medline: [40805134](https://pubmed.ncbi.nlm.nih.gov/40805134/)]
58. Ma DD, Liu Z, Au K, et al. Randomized controlled trial of a virtually delivered exercise and stress management program to improve physical performance of hematopoietic cell transplant survivors. *J Clin Oncol*. Mar 10, 2025;43(8):949-959. [doi: [10.1200/JCO.24.00333](https://doi.org/10.1200/JCO.24.00333)] [Medline: [39591546](https://pubmed.ncbi.nlm.nih.gov/39591546/)]
59. Unick JL, Duffy C, Dizon D, et al. Evaluation of a translatable web-based intervention for increasing physical activity among cancer survivors: pilot randomized trial. *JMIR Cancer*. Oct 2, 2025;11:e79610. [doi: [10.2196/79610](https://doi.org/10.2196/79610)] [Medline: [41037739](https://pubmed.ncbi.nlm.nih.gov/41037739/)]
60. Yang YH, Chao YL, Lin YF, Liu PC, Chang KJ, Hou IC. Home-based remote dance program with biopsychosocial model improves quality of life in breast cancer patients: a randomized controlled trial. *J Bodyw Mov Ther*. Sep 2025;43:376-385. [doi: [10.1016/j.jbmt.2025.04.029](https://doi.org/10.1016/j.jbmt.2025.04.029)] [Medline: [40483151](https://pubmed.ncbi.nlm.nih.gov/40483151/)]
61. Yu K, Yin B, Zhu Y, et al. Efficacy of a digital postoperative rehabilitation intervention in patients with primary liver cancer: randomized controlled trial. *JMIR mHealth uHealth*. Apr 7, 2025;13:e59228. [doi: [10.2196/59228](https://doi.org/10.2196/59228)] [Medline: [40194311](https://pubmed.ncbi.nlm.nih.gov/40194311/)]
62. Zhou R, Chen Z, Zhang S, et al. Effects of exercise on cancer-related fatigue in breast cancer patients: a systematic review and meta-analysis of randomized controlled trials. *Life (Basel)*. Aug 14, 2024;14(8):1011. [doi: [10.3390/life14081011](https://doi.org/10.3390/life14081011)] [Medline: [39202753](https://pubmed.ncbi.nlm.nih.gov/39202753/)]
63. Wang Y, Yang L, Lin G, et al. The efficacy of progressive muscle relaxation training on cancer-related fatigue and quality of life in patients with cancer: a systematic review and meta-analysis of randomized controlled studies. *Int J Nurs Stud*. Apr 2024;152:104694. [doi: [10.1016/j.ijnurstu.2024.104694](https://doi.org/10.1016/j.ijnurstu.2024.104694)] [Medline: [38281450](https://pubmed.ncbi.nlm.nih.gov/38281450/)]
64. Li J, Zhu C, Liu C, Su Y, Peng X, Hu X. Effectiveness of eHealth interventions for cancer-related pain, fatigue, and sleep disorders in cancer survivors: a systematic review and meta-analysis of randomized controlled trials. *J Nurs Scholarsh*. Mar 2022;54(2):184-190. [doi: [10.1111/jnu.12729](https://doi.org/10.1111/jnu.12729)] [Medline: [34791779](https://pubmed.ncbi.nlm.nih.gov/34791779/)]
65. de Sousa D, Fogel A, Azevedo J, Padrão P. The effectiveness of web-based interventions to promote health behaviour change in adolescents: a systematic review. *Nutrients*. Mar 16, 2022;14(6):1258. [doi: [10.3390/nu14061258](https://doi.org/10.3390/nu14061258)] [Medline: [35334915](https://pubmed.ncbi.nlm.nih.gov/35334915/)]
66. Wang Z, Li Y, Wang Q, Su Y. The effectiveness of wearable electronic device system-supported physical activity programs for cancer survivors: meta-analysis of randomized controlled trials. *J Med Internet Res*. Aug 14, 2025;27:e74347. [doi: [10.2196/74347](https://doi.org/10.2196/74347)] [Medline: [40811695](https://pubmed.ncbi.nlm.nih.gov/40811695/)]
67. Bower JE, Wiley J, Petersen L, Irwin MR, Cole SW, Ganz PA. Fatigue after breast cancer treatment: biobehavioral predictors of fatigue trajectories. *Health Psychol*. Nov 2018;37(11):1025-1034. [doi: [10.1037/hea0000652](https://doi.org/10.1037/hea0000652)] [Medline: [30321021](https://pubmed.ncbi.nlm.nih.gov/30321021/)]
68. Shaffer KM, Turner KL, Siwik C, et al. Digital health and telehealth in cancer care: a scoping review of reviews. *Lancet Digit Health*. May 2023;5(5):e316-e327. [doi: [10.1016/S2589-7500\(23\)00049-3](https://doi.org/10.1016/S2589-7500(23)00049-3)] [Medline: [37100545](https://pubmed.ncbi.nlm.nih.gov/37100545/)]
69. Filis P, Markozannes G, Chan DS, et al. Grading the evidence for physical activity and any outcome in cancer survivors: an umbrella review of 740 meta-analytic associations. *Crit Rev Oncol Hematol*. Mar 2025;207:104602. [doi: [10.1016/j.critrevonc.2024.104602](https://doi.org/10.1016/j.critrevonc.2024.104602)] [Medline: [39730034](https://pubmed.ncbi.nlm.nih.gov/39730034/)]
70. Choi E, Ryu S, Chun H, Kwak W, Choi SK. Personal competence factors associated with better access to digital health [Report in Korean]. Korea Institute for Health and Social Affairs; 2022. URL: <https://www.kihasa.re.kr/en/publish/paper/research/view?seq=46808> [Accessed 2025-06-08]

Abbreviations

CRF: cancer-related fatigue

DHI: digital health intervention

EORTC QLQ C-30: European for Research and Treatment of Cancer Quality of Life Questionnaire C-30

ESF: Evidence Standards Framework

FACIT: Functional Assessment of Chronic Illness Therapy

JBI: Joanna Briggs Institute

MESH: Medical Subject Headings

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

PRISMA-ScR: Preferred Reporting Items for Systematic Reviews and Meta-Analyses extension for Scoping Review

PROSPERO: International Prospective Register of Systematic Reviews

RCTs: randomized controlled trials

Edited by Stefano Brini; peer-reviewed by Georgios Grigoriadis, Yufang Hao; submitted 07.Sep.2025; final revised version received 14.May.2026; accepted 15.May.2026; published 26.Jun.2026

Please cite as:

Kim Y, Bae KR, Sung JH, Ko YH, Kim S

Physical Activity Interventions Using Digital Health Interventions for Cancer-Related Fatigue in People With a History of Cancer: Scoping Review

J Med Internet Res 2026;28:e83727

URL: <https://www.jmir.org/2026/1/e83727>

doi: [10.2196/83727](https://doi.org/10.2196/83727)

© Yeeun Kim, Ka Ryeong Bae, Ji Hyun Sung, Yun Hee Ko, Sue Kim. Originally published in the Journal of Medical Internet Research (<https://www.jmir.org>), 26.Jun.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.