Review

Effectiveness of Digital Health Interventions in Promoting Physical Activity Among College Students: Systematic Review and Meta-Analysis

Siyuan Bi, MA; Junfeng Yuan, MA; Yanling Wang, MA; Wenxin Zhang, MA; Luqin Zhang, MA; Yongjuan Zhang, MA; Rui Zhu, MA; Lin Luo, PhD

School of Physical Education, Guizhou Normal University, Guiyang, China

Corresponding Author:

Lin Luo, PhD School of Physical Education Guizhou Normal University Huaxi District, University Town, Siya Road Guiyang, 550025 China Phone: 86 86751983 Email: <u>460022831@gznu.edu.cn</u>

Abstract

Background: Recent studies offer conflicting conclusions about the effectiveness of digital health interventions in changing physical activity behaviors. In addition, research focusing on digital health interventions for college students remains relatively scarce.

Objective: This study aims to examine the impact of digital health interventions on physical activity behaviors among college students, using objective measures as outcome indicators.

Methods: In accordance with the 2020 PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analyses) guidelines, a comprehensive literature search was conducted across several databases, including MEDLINE (PubMed), Web of Science, Cochrane Library, and EBSCO (CINAHL Plus with full text), to identify relevant intervention studies published up to June 6, 2023. The inclusion criteria specified studies that examined the quantitative relationships between digital health interventions and physical activity among adults aged 18 years to 29 years, focusing on light physical activity (LPA), moderate to vigorous physical activity (MVPA), sedentary time (ST), or steps. Non–randomized controlled trials were excluded. The quality of the studies was assessed using the Cochrane Risk of Bias tool. Results were synthesized both narratively and quantitatively, where applicable. When sufficient homogeneity was found among studies, a random-effects model was used for meta-analysis to account for variability.

Results: In total, 8 studies, encompassing 569 participants, were included in the analysis. The primary outcomes measured were LPA, MVPA, ST, and steps. Among these studies, 3 reported on LPA, 5 on MVPA, 5 on ST, and 3 on steps. The meta-analysis revealed a significant increase in steps for the intervention group compared with the control group (standardized mean difference [SMD] 0.64, 95% CI 0.37-0.92; P<.001). However, no significant differences were observed between the intervention and control groups regarding LPA (SMD -0.08, 95% CI -0.32 to 0.16; P=.51), MVPA (SMD 0.02, 95% CI -0.19 to 0.22; P=.88), and ST (SMD 0.03, 95% CI -0.18 to 0.24; P=.78).

Conclusions: Digital health interventions are effective in increasing steps among college students; however, their effects on LPA, MVPA, and sedentary behavior are limited.

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KEYWORDS

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digital health intervention; college students; physical activity behavior; steps; light intensity physical activity; moderate to vigorous physical activity; sedentary behavior; knowledge synthesis; systematic review; meta-analysis; physical activity; eHealth; digital health; mobile phone

Introduction

Background

The college years represent a critical phase [1] that significantly influences the levels of physical activity and the formation of healthy habits [2-4]. However, research indicates that physical activity levels among college students are generally low and rarely meet the recommended health guidelines [5]. Physical activity and sedentary behavior are interrelated behaviors that affect health and well-being and are risk factors for noncommunicable diseases. The increasing number of sedentary adults negatively impacts the prevalence of noncommunicable diseases and the overall health status of the global population [6-9]. Physical activity is a crucial protective factor that can prevent and manage major noncommunicable diseases such as cardiovascular diseases, stroke, diabetes, breast cancer, and colon cancer. Furthermore, physical activity helps prevent other significant risk factors for noncommunicable diseases, such as hypertension, overweight, and obesity, while also improving mental health, delaying the onset of dementia, and enhancing the quality of life and well-being. Based on this, the World Health Organization has developed global recommendations for promoting health through physical activity, suggesting that adults should engage in at least 150 minutes of moderate-intensity physical activity or 75 minutes of vigorous-intensity physical activity per week. However, currently, a quarter of adults worldwide fail to meet the global physical activity recommendations as set by the World Health Organization [10]. Thus, efforts to increase physical activity and reduce sedentary behavior have been identified as international public health priorities [11-13].

The lack of physical activity negatively affects the health of college students, potentially leading to psychological health issues, academic performance decline, and limited social life [14-16]. To improve college students' physical activity behavior, digital health interventions have been recognized as a promising strategy [17,18]. Digital health is a health strategy that uses digital, mobile, and wireless technologies to support the achievement of health goals, including mobile health (mHealth) and eHealth. These technologies encompass mobile phones, tablets, web-based interventions, SMS, social media, fitness mobile apps, and wearable devices [19]. Digital health interventions may involve the use of mobile applications, social media platforms, online fitness programs, and other digital tools to provide personalized health advice, goal setting, and monitoring [20-23]. A characteristic feature of digital health interventions is their ease of deployment and the ability to target specific population groups [19,24]. The Global Action Plan on Physical Activity by the World Health Organization emphasizes the use of digital health interventions to promote physical activity behavior [25]. The advantages of digital health interventions lie in their reliability, real-time nature, and personalized customization, which can provide effective support for physical activity among college students [26-33]. Considering that college students are characterized as digital natives [34-36], scholars have conducted numerous digital health intervention studies targeting college students' health behaviors. However, despite the extensive research on digital health

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interventions, the overall effectiveness of these interventions on college students' physical activity behavior remains unclear.

Some studies have found that personalized intelligent SMS interventions have a positive impact on physical activity health outcomes [37,38]. Other research supports the effectiveness of smartphone applications in increasing physical activity [39-44]. However, there are also research findings that suggest digital health interventions may not effectively promote changes in physical activity behavior [45-47]. In addition, some studies have found that internet-based interventions have positive effects on physical activity, but with small effect sizes and heterogeneity [48,49]. Furthermore, other studies suggest that digital health interventions may improve specific health behaviors and outcomes, but with limited effectiveness [50,51]. There are also studies that indicate a small positive correlation between digital health engagement and physical activity, but the heterogeneity in the design, settings, intervention components, and outcomes of the studies may affect the validity of the conclusions [52-54]. In summary, while digital health interventions have the potential to improve college students' physical activity behavior, further research is needed to validate their overall effectiveness.

This study focused on young adults aged 18-29 years, specifically college students in the early adulthood stage [1]. This demographic was chosen due to its unique developmental phase, characterized by a transition from adolescence to adulthood. Individuals in this group face numerous challenges, including academic pressure, changing living environments, and shifting social roles, all of which can contribute to decreased physical activity levels and an increased risk of obesity and related chronic diseases [2-4].

Digital health interventions, as a convenient, low-cost, and easily disseminated health promotion tool, have shown potential in promoting physical activity in recent years [19,24]. However, existing research presents mixed findings regarding the effectiveness of such interventions on college students' physical activity. While some studies report positive impacts on physical activity indicators [37,38], others suggest limited or even insignificant effects [45-47].

To address this ambiguity, a meta-analysis was conducted to synthesize and analyze existing research findings. This rigorous approach aimed to provide more comprehensive and reliable evidence for evaluating the impact of digital health interventions on college students' physical activity behavior. The findings contributed to filling the knowledge gaps in the current literature and offered empirical evidence to support the application of digital health interventions in promoting physical activity among this population.

Objectives

The aim of this study is to assess the overall impact of digital health interventions on college students' physical activity behavior through a meta-analysis. By synthesizing and summarizing existing research findings, the study seeks to obtain an accurate evaluation of the overall effectiveness of digital health interventions in the college student population. In this context, the following research questions are proposed: Do

digital health interventions have a positive impact on college students' physical activity behavior? Can digital health interventions significantly improve the level of physical activity among college students? Are there differences in the effectiveness of digital health interventions across different modes of physical activity?

By addressing these research questions, the study aims to develop a comprehensive understanding of the impact of digital health interventions on college students' physical activity behavior. The findings are intended to provide scientific evidence to support the development of relevant policies and intervention measures.

Methods

Overview

This systematic review was conducted in accordance with the PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analyses) 2020 guidelines [55] and was registered with PROSPERO (International Prospective Register of Systematic Reviews) under the identifier CRD42024533180. A rigorous and comprehensive literature search was executed

using computer-assisted systematic search strategies. Initially, as specified in our registered protocol (CRD42024533180), the intention was to limit the search to the MEDLINE (PubMed) and Web of Science databases. However, to ensure a more comprehensive inclusion of relevant literature, the search strategy was subsequently expanded to incorporate additional databases, namely the Cochrane Library and EBSCO (CINAHL Plus with full text). Apart from this modification in the search strategy, the study design remained consistent with the original PROSPERO protocol.

Search Strategy

This study aimed to assess the impact of digital health interventions on changes in physical activity behavior among college students. To ensure a comprehensive search, we searched MEDLINE (PubMed) and Web of Science, Cochrane library, and EBSCO (CINAHL Plus with full text) databases for relevant articles published up to June 6, 2023. The detailed search strategy, including specific keywords and search terms, is available in Multimedia Appendix 1.

Selection Criteria

The selection criteria are shown in Textbox 1.

Textbox 1. List of inclusion and exclusion criteria.

Inclusion criteria

- Population: the study population consists of college students aged 18-29 years. Participants with intellectual disabilities, significant cognitive impairments, or severe mobility impairments were excluded from this study.
- Interventions: the included interventions in the study are digital health-based interventions that have an impact on changes in physical activity behavior. These interventions include mHealth and eHealth, such as mobile phones, tablets (such as iPad), web-based interventions (such as web-based chats and teleconferences), SMS, social media, lifestyle or fitness smartphone apps, and wearable devices.
- Control group: the included control groups in the study consist of groups that do not receive any intervention, receive traditional interventions, or receive minimal interventions (such as providing only physical activity goals). If there are multiple intervention groups in the study, the control group's data from and the data from the intervention group with the most digital health intervention features are extracted for analysis.
- Outcomes: this meta-analysis includes studies that measure physical activity or sedentary behavior objectively using devices (such as mobile phone data, accelerometers, pedometers). The mean and SD of the following indicators are extracted: light physical activity and moderate to vigorous physical activity, sedentary time, and step count.
- Study design: this meta-analysis only includes randomized controlled trials, where the experimental group uses technology-based interventions while the control group receives either traditional interventions (primarily oral or written health education materials) or no intervention.

Exclusion criteria

- Review article.
- Letter to the editor.
- Editorial.
- Articles in non-English languages (while acknowledging that excluding non-English articles may introduce language bias and limit the comprehensiveness of the findings, this review focuses solely on research published in English due to resource limitations, specifically regarding translation, time, and funding).
- Other study designs (retrospective studies, quasi-experimental designs, non-randomized controlled trials).
- Articles based on studies focused on research subjects other than college students (aged 18-29 years).
- Articles that did not collect or report data on sex.
- Articles without objective measurement of physical activity (eg, relying solely on self-report).
- Articles exclusively reporting on the feasibility or acceptability of technology-based interventions, or lacking findings pertaining to physical activity behavior outcomes, were excluded.
- Articles with incomplete outcome reporting, hindering reliable synthesis, were excluded from this analysis.

Study Selection

To guarantee the accuracy and reliability of included studies, a rigorous screening process was implemented. Following the guidelines for meta-analysis, 2 independent reviewers conducted a blinded, duplicate screening of all retrieved articles to determine their eligibility for inclusion [56]. The screening process involved 2 stages. Initially, titles and abstracts were screened to exclude studies that clearly did not meet the inclusion criteria. Subsequently, full texts of the remaining articles were reviewed to confirm their eligibility. Throughout the screening process, discrepancies between reviewers were resolved through discussion and consensus.

Data Extraction

The researchers used standardized forms specifically developed for this review to independently extract data from each included study, ensuring comprehensive acquisition of trial data. The extracted data encompassed the study country, study participants (population and sample size), study design (description of digital intervention measures, intervention and control group characteristics, intervention and duration), outcome measurements (measurement tools and timing), and key study outcomes (mean and SD of light physical activity [LPA], moderate to vigorous physical activity [MVPA], sedentary time [ST], and steps, or baseline and follow-up step count statistics). Average changes were computed when necessary, and in cases where SDs were not reported, they were calculated based on SE or 95% CI [57,58]. If the research papers did not adequately report the results, the researchers contacted the authors to obtain additional data. In instances where discrepancies in data extraction arose, the author team engaged in discussions and reexamined the original studies until a consensus was reached.

Risk of Bias (Quality) Assessment

The researchers used the Cochrane Collaboration bias tool, which is integrated within the Review Manager software (version 5.4, The Nordic Cochrane Centre), and supplemented it with the CONSORT (Consolidated Standards of Reporting Trials) checklist to assess the quality of the included studies [59]. This tool comprises domains for selection bias, performance bias, detection bias, attrition bias, and reporting bias. Each study was independently assessed by 2 reviewers. In addition, publication bias was assessed using funnel plot analysis.

Data Analysis

The statistical analysis for this study was performed using Review Manager and Stata (version 16; StataCorp LLC). Considering the challenges posed by various definitions of changes in physical activity behavior, this study decided to include only literature that solely used LPA, MVPA, ST, and steps as pre- and postintervention measurement indicators. Studies that did not provide these measurement data were excluded to ensure interpretability and meaningfulness of the meta-analysis. In addition, studies that did not provide sufficient information for calculating the mean and SD, or used statistical analyses (such as correlation coefficients, multivariate, and adjusted regression coefficients) inconsistent with the relevant meta-analysis, were also excluded. Furthermore, in cases where multiple intervention groups within a study exhibited similar characteristics of digital health interventions, the intervention group with the most digital health intervention features was selected for the meta-analysis.

The primary outcome measures for this meta-analysis were the mean and SD of LPA, MVPA, ST, and steps. For each intervention and control group in each study, the baseline and follow-up mean, SD, and participant count of physical activity were entered into the Review Manager software. Using these data, standardized mean differences (SMDs) between intervention and control group changes for each study were calculated as the measure of effect size. The SMD was chosen because although the included studies used the same units of measurement, the scales or measurement tools used across studies might have varied, making direct comparisons of weighted mean differences difficult. SMD, through standardization, can eliminate the influence of different scales, reflecting the differences in intervention effects more accurately and enhancing the interpretability of the results. If data were available from 3 or more studies, a meta-analysis was conducted, and an overall effect size was computed.

Due to the diversity in study populations and intervention designs, a random effects model approach was used to account for heterogeneity among studies. The P statistic was used to assess the presence of heterogeneity, which describes the percentage of total variation in effect estimates due to heterogeneity [60,61]. The P statistic was chosen as the preferred measure of variance since it is stable with small sample sizes. Heterogeneity was considered significant if the P value from the chi-square test was less than .10 and P was greater than 50% [60,62,63].

The effect size of SMDs was interpreted using Cohen suggestion [64], where 0.2 represents a small effect size, 0.5 represents a medium effect, and 0.8 represents a large effect, with values below 0.2 considered trivial. In addition, sensitivity analyses for LPA, MVPA, ST, and steps were performed using Stata software (version 16.0) to further assess the stability of the study results.

Results

Study Selection

Initially, a total of 14,915 articles were identified for evaluation across 4 databases, such as 1743 from PubMed, 6403 from Web of Science, 4583 from the Cochrane Library, and 1933 from EBSCO (CINAHL Plus with full text). In addition, 253 articles were identified through manual searching of reference lists from key articles. After excluding 4028 duplicate records from multiple databases and search channels, we proceeded to exclude 10,722 articles that were irrelevant to the scope of this study based on their titles and abstracts. From the remaining 165 articles, an additional 157 were deemed ineligible for inclusion in this meta-analysis, as they did not meet the predetermined inclusion criteria. Ultimately, a total of 8 relevant research articles were identified and included in this meta-analysis. The detailed flowchart of the literature search process is presented in Figure 1.

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Figure 1. PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analyses) diagram of the searching and screening processes.



Study Characteristics

Overview

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A total of 8 studies were included in this meta-analysis, all of which were randomized controlled trials published in English between 2014 and 2023. The main characteristics of these 8 studies are provided in Multimedia Appendix 2 (PRISMA checklist provided in Multimedia Appendix 3). It was reported that four studies [45,46,50,65] were not registered on

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ClinicalTrials.gov, while 4 studies [47,51,66,67] were registered on the website. Among these 8 studies, 4 were conducted in Australia [45], Spain [66], China [46], and Germany [65], respectively, while the remaining 4 studies were conducted in the United States [47,50,51,67].

Participants

The sample sizes of these 8 studies ranged from 34 to 187 participants, with a total of 569 participants. All included studies

targeted individuals between the ages of 18 and 30 years, specifically university students. Among these, except for 1 study [50] which did not report or calculate the SD of participant age, the remaining 7 studies [45-47,51,65-67] reported or calculated both the mean and SD of participant age.

The 8 studies exhibited varying gender distributions among their participants. In a study by Hebden et al [45], women comprised 80% (n=41) of the 51 participants. Similarly, a study by Kim et al [47] included 187 participants with 62% (n=116) identifying as female. The study by Miragall et al [66] comprised 52 participants, of whom 86% (n=45) were women. A study by Pope at al [51], with 38 participants, had 74% (n=28) female representation. The study by Al-Nawaiseh et al [67] included 114 participants, with women constituting 81% (n=92) of the sample. The study by Lau et al [46] had a more balanced gender distribution, with 61% (n=34) of its 56 participants being female. The study by Pope and Gao [50], involving 44 participants, had 73% (n=32) female participants. Finally, the study by Kellner et al [65], with the smallest sample size of 34, had the highest female representation at 85% (n=29).

Interventions

All primary interventions in the studies aimed to increase physical activity, although one study [45] aimed to evaluate the impact on body weight, and another study [50] aimed to assess the impact on health outcomes. The methods used in the intervention group included social media [46], smart text messaging [65], mobile digital applications [45,67], smart wearable technology devices (such as smartwatches and pedometers) [47,66], or a combination of social media and

smartphone apps [50], or a combination of social media and wearable technology devices [51]. The control group either did not receive any intervention [46,65] or only received traditional/standard interventions [45,47,50,51,66,67]. The duration of the interventions varied from 3 weeks [66] to 4 weeks [46], 5 weeks [65], 10 weeks [50], 12 weeks [45,51,67], and 15 weeks [47].

Outcomes

All included studies reported multiple objective measures of physical activity. In total, 3 studies (n=276) reported LPA, 5 studies (n=369) reported MVPA, 5 studies (n=354) reported ST, and 3 studies (n=215) reported steps.

For measuring LPA, MVPA, and ST, the 8 studies used the following tools: ActiGraph GT3X+ accelerometer [68], ActiGraph accelerometer (GT1M model) [69], SenseWear Pro Armband mobile device [70], ActiGraph Actitrainer (ActiGraph LLC) [71], ActiGraph Link accelerometer [72], and ActivPal accelerometer (Pal Technologies) [73]. The tools used to measure steps included the Pacer pedometer, Accupedo pedometer, and Fitbit One pedometer.

Risk of Bias in Included Studies

The methodological rigor of the included studies was evaluated using the Cochrane Risk of Bias tool, which assesses 6 domains, namely selection bias, performance bias, detection bias, attrition bias, reporting bias, and other biases. Each domain was assessed as having a low risk, high risk, or unclear risk of bias. The overall quality of each study was then classified as high, moderate, or low based on the assessment results (Figure 2 [45-47,50,51,65-67]).

Figure 2. Risk of bias summary. These are authors' judgments of each methodological quality item for each included study. Plus signs (+) indicate high methodological quality (low risk of bias); minus signs (-) indicate low methodological quality (high risk of bias); question marks (?) indicate unclear methodological quality (reported information about what happened in the study was insufficient).



The included studies generally demonstrated moderate methodological quality, meeting average quality assessment standards. Notably, all but one study [46] explicitly described a robust random sequence generation process for participant allocation, minimizing selection bias. Furthermore, all studies reported baseline characteristics and used valid, objective

measures for assessing physical activity outcomes, reducing the risk of detection bias.

However, the nature of the interventions presented challenges in achieving blinding of participants and personnel, potentially introducing performance bias. Specifically, while intention-to-treat analysis was used in all but 4 studies

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[46,65-67], mitigating attrition bias, the lack of blinding may have influenced participant behavior and data interpretation.

Regarding reporting bias, all studies, with the exception of Lau et al [46], provided sufficient statistical data for analysis. However, only 5 studies [46,47,50,51,67] reported between-group differences in changes over time, potentially limiting the comprehensiveness of the findings.

Meta-Analysis

Figure 3 displays the funnel plots for these measures, assessing publication bias and heterogeneity. The plots for LPA and steps

indicate low publication bias and heterogeneity, with data points symmetrically distributed around the funnel's center, suggesting consistent effect sizes. In contrast, the MVPA plot shows higher heterogeneity and potential publication bias, with data points more dispersed and deviating from the center line. The ST plot, while having some points at the funnel's edges, overall indicates slight publication bias and low heterogeneity, with generally consistent effect sizes. These findings highlight the need for researchers to consider heterogeneity and potential publication bias when interpreting results for different physical activity types and sedentary time (ST).

Figure 3. Funnel plot of light physical activity (LPA), moderate to vigorous physical activity (MVPA), sedentary time (ST), and steps.



The results of the meta-analysis showed the forest plot (Figure 4 [45-47,50,51,65-67]) for the 4 outcome measures: LPA, MVPA, ST, and steps.



Figure 4. Forest plot of light physical activity (LPA), moderate to vigorous physical activity (MVPA), sedentary time (ST), and steps per week [45-47,50,51,65-67].

	Intervention			Control				Std. Mean Difference	Std. Mean Difference
Study or Subgroup	Mean	SD	Total	Mean	SD	Total	Weight	IV, Random, 95% Cl	IV, Random, 95% Cl
1.1.1 LPA									
Hebden et al [45], 2014	1,671.6	354.2	26	1,579.9	297.5	25	19.0%	0.28 [-0.28, 0.83]	
Kim et al [47], 2018	1,260.49	825.93	101	1,380.68	763.42	86	66.8%	-0.15 [-0.44, 0.14]	
Pope et al [51], 2019	1,149.4	301.7	19	1,228.5	358.4	19	14.2%	-0.23 [-0.87, 0.40]	
Subtotal (95% CI)			146			130	100.0%	-0.08 [-0.32, 0.16]	-
Heterogeneity: Tau² = 0.00	; Chi ² = 2.04,	df = 2 (P = 0	0.36); I ^z	= 2%					
Test for overall effect: Z = 0	.66 (P = 0.51)							
1.1.2 MVPA									
Hebden et al [45], 2014	298.2	180.6	26	270.2	112.7	25	13.8%	0.18 [-0.37, 0.73]	
Kim et al [47], 2018	351.26	311.64	101	351.75	288.19	86	50.6%	-0.00 [-0.29, 0.29]	
Lau et al [46], 2022	401.94	136.99	24	408.94	172.13	25	13.3%	-0.04 [-0.60, 0.52]	
Pope and Gao (50), 2022	260.4	87.5	22	285.6	128.8	22	11.9%	-0.22 [-0.82, 0.37]	
Pope et al [51], 2019	56.7	39.2	19	46.9	42.7	19	10.3%	0.23 [-0.40, 0.87]	
Subtotal (95% CI)			192			177	100.0%	0.02 [-0.19, 0.22]	-
Heterogeneity: Tau ² = 0.00	; Chi² = 1.49,	df = 4 (P = 0	0.83); I²	= 0%					
Test for overall effect: Z = 0	.15 (P = 0.88)							
1.1.3 ST									
Hebden et al [45], 2014	3,710	686	26	3,848.6	446.6	25	14.4%	-0.23 [-0.79, 0.32]	
Kellner et al [65], 2023	3,685.99	558.11	16	3,908.66	888.16	18	9.5%	-0.29 [-0.97, 0.39]	
Kim et al [47], 2018	3,817.8	928.2	101	3,696	777	86	52.8%	0.14 [-0.15, 0.43]	-+ -
Pope and Gao (50), 2022	3,017	504.7	22	3,031.7	272.3	22	12.5%	-0.04 [-0.63, 0.56]	
Pope et al [51], 2019	3,840.9	312.2	19	3,772.3	352.1	19	10.8%	0.20 [-0.44, 0.84]	
Subtotal (95% CI)			184			170	100.0%	0.03 [-0.18, 0.24]	
Heterogeneity: Tau ² = 0.00	; Chi ² = 2.63,	df = 4 (P = 0	0.62); I ^z	= 0%					
i est for overall effect: $Z = U$	28 (P = 0.78	9							
1.1.4 Steps									
Al-Nawaiseh[67], 2022	54,896.27	14,992.35	56	45,530.12	15,703.41	58	53.6%	0.61 [0.23, 0.98]	
Lau et al [46], 2022	65,562	17,430	24	56,728	23,107	25	23.6%	0.42 [-0.14, 0.99]	
Miragall et al [66], 2018	55,706	14,035	26	43,757	10,388	26	22.8%	0.95 [0.38, 1.53]	
Subtotal (95% CI)			106	~~		109	100.0%	0.64 [0.37, 0.92]	
Heterogeneity: Tau ² = 0.00	; Chi* = 1.73,	at = 2 (P = 0	J.42); I²	= 0%					
lest for overall effect: Z = 4	.57 (P < 0.00	001)							
								_	
									-1 -0.5 0 0.5 1

Test for subaroup differences: Chi² = 18.12. df = 3 (P = 0.0004), i^2 = 83.4%

Regarding LPA, 3 studies involving 276 participants reported changes in weekly LPA. The results revealed a small SMD between the intervention and control groups, slightly favoring the control group (SMD –0.08, 95% CI –0.32 to 0.16; Figure 4 [45-47,50,51,65-67]). However, this difference was not statistically significant (P=.51). In addition, the results showed very low heterogeneity (P=2%).

For MVPA, 5 studies involving 369 participants reported changes in weekly MVPA. The results indicated a small SMD between the intervention and control groups, slightly favoring the intervention group (SMD 0.02, 95% CI –0.19 to 0.22; Figure 4 [45-47,50,51,65-67]). However, this difference was not statistically significant (P=.88). Furthermore, the results showed no heterogeneity (P=0%).

Regarding ST, 5 studies involving 354 participants reported changes in weekly ST. The results showed a small SMD between the intervention and control groups, slightly favoring the intervention group (SMD 0.03, 95% CI –0.18 to 0.24; Figure 4 [45-47,50,51,65-67]). However, this difference was not statistically significant (P=.78). In addition, the results showed no heterogeneity (P=0%).

For steps, 3 studies involving 215 participants reported changes in weekly step counts. The results revealed a large SMD between the intervention and control groups, significantly favoring the intervention group (SMD 0.64, 95% CI 0.37-0.92; Figure 4 [45-47,50,51,65-67]), with a statistically significant difference (P<.001). Furthermore, the results showed no heterogeneity (P=0%).

Favours [Intervention] Favours [Control]

Sensitivity Analysis

The research findings of the 4 outcome measures, namely LPA, MVPA, ST, and steps, indicate that the point estimates of the combined effect sizes obtained after excluding a specific study fall within the 95% CI of the overall combined effect size. This suggests that the exclusion of a particular study does not significantly alter the results, indicating a considerable stability in the evaluation of the 4 outcome measures. The sensitivity analysis charts corresponding to each outcome measure are presented in figures.

The sensitivity analysis chart for LPA is shown in Figure 5 [45,47,51].



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Figure 5. Sensitivity analysis chart of light physical activity (LPA) [45,47,51].



The vertical line labeled "Estimate" represents the combined results of all included studies, while the lines on either side denote the upper and lower confidence intervals of these results. A circle corresponding to any particular study indicates the combined results of the remaining studies after excluding that specific study. If, upon exclusion of a study, the circle remains close to the "Estimate" line and within the original confidence interval, it suggests that the exclusion of this study does not significantly alter the combined results. This consistency implies that the meta-analysis results are robust.

The sensitivity analysis charts for MVPA, ST, and steps are shown in Figures 6 [45-47,50,51], 7 [45,47,50,51,65], and 8 [46,66,67], respectively.



Meta-analysis estimates, given named study is omitted

Lower CI Limit
Estimate
Upper CI Limit

Hebden 2014
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Figure 6. Sensitivity analysis chart of moderate to vigorous physical activity (MVPA) [45-47,50,51].

Figure 7. Sensitivity analysis chart of sedentary time (ST) [45,47,50,51,65].





Figure 8. Sensitivity analysis chart of steps [46,66,67].



Discussion

Principal Findings

The results of the included analysis demonstrate that digital health interventions are a stable and effective strategy for significantly improving the step count of college students, thereby enhancing their physical health. This conclusion is supported by 3 randomized controlled trials included in the analysis [46,66,67]. With the widespread use of health trackers and mobile devices, daily step count has become a convenient and reliable indicator for measuring individual physical activity levels [74,75]. Research indicates that increasing daily step count has significant benefits for both physical and mental health [76]. First, increasing step count enhances cardiopulmonary function and oxygen uptake, thereby improving cardiovascular function [75,77,78]. Second, walking is a low-intensity aerobic exercise that can help lower blood pressure, control blood sugar levels, and prevent cardiovascular diseases. In addition, walking aids in calorie expenditure, weight control, and reduces the risk of obesity. Furthermore, walking helps release physical and psychological stress, improving emotional and mental well-being [79-81]. Further studies have demonstrated a correlation between daily step count and all-cause mortality. Individuals who walk 8000 to 12,000 steps per day have a lower all-cause mortality rate compared with those who walk only 4000 steps per day. This association has been validated across different age groups, genders, and ethnicities [82,83]. Therefore, focusing on step count and increasing it through intervention measures is of significant importance. Digital health interventions provide a convenient and practical approach to help college students

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increase their step count. By using digital health technologies such as the internet, smartphone apps, text messaging services, and wearable devices, individuals can monitor their step count, set goals, receive reminders and feedback, and more. This intervention not only alerts individuals to their walking behavior but also provides personalized advice and support to encourage increased step count.

However, the analysis results of this study indicate that digital health interventions may not be an effective method for significantly influencing the levels of low-, moderate-, and high-intensity physical activity and ST among college students. Although previous research has demonstrated that increasing the duration of low, moderate, and high-intensity physical activity can significantly reduce the negative impact of ST and improve risk factors associated with cardiovascular diseases, such as hypertension, hyperglycemia, hyperlipidemia, and mortality [84-87]. However, in this study, no significant impact of digital health interventions on the levels of low, moderate, and high-intensity physical activity and ST among college students was observed. It is worth noting that in this study, the impact of digital health interventions on step count was significant, but not significant for other modes of physical activity. This may be due to the fact that walking is a low-intensity aerobic exercise that is easily accepted and maintained by college students. In contrast, changing high-intensity physical activity and sedentary behaviors may require more personalized support and intervention measures.

Specifically, step count differs from LPA measured by accelerometers, and thus, the significance of step count does not represent the significance of LPA. Loprinzi and Cardinal [88] found that while an increase in step count is significantly associated with health, the correlation between step count and total LPA is low, indicating that the significance of step count is not equivalent to LPA. Tudor-Locke et al [89] also noted that step count primarily measures walking, whereas LPA encompasses a broader range of low-intensity activities.

In previous studies, the impact on LPA and step count yielded different results. Hebden [45] found that LPA levels significantly increased in both the intervention and control groups, with no significant difference between them. Kim et al [47] reported a significant decrease in LPA time in the intervention group, with no significant change in the control group, yet no significant difference between the groups. Pope et al [51] observed a decrease in LPA time in the intervention group and no significant change in the control group, with no significant difference between the groups. Regarding step count, Miragall et al [66] found a significant increase in daily steps in the intervention group, with no significant change in the control group, and a significant difference between the groups. Lau et al [46] found no significant difference in objectively measured step count between the intervention and control groups. Al-Nawaiseh et al [67] observed a significant increase in weekly steps from baseline to follow-up in the intervention group, with no significant change in the control group, and a significant difference between the groups.

Therefore, to effectively promote low-, moderate-, and high-intensity physical activity among college students and improve their sedentary behaviors, it is necessary to explore more effective methods and continue this investigation in future research. This may involve providing more personalized intervention measures, such as customized exercise plans, individual goal setting, and regular feedback. In addition, combining social support and motivational mechanisms, such as team activities and competitions, may further encourage college students to engage in more high-intensity physical activity and reduce sedentary behaviors.

In light of the above, digital health interventions have been identified as an effective strategy to enhance step count and improve the physical health status of college students. By monitoring step count, setting goals, and providing personalized advice and support, digital health interventions can assist college students in increasing their daily step count. However, further research is needed to investigate the effectiveness of interventions targeting other modes of physical activity and to delve into the mechanisms underlying digital health interventions. With the continuous advancement of digital health technologies, future studies should explore more precise and personalized intervention approaches and conduct in-depth investigations into the mechanisms of digital health interventions to provide a more comprehensive and profound understanding.

Strengths and Limitations

The included studies in this meta-analysis used various intervention measures, including eHealth and mHealth technologies such as the internet, smartphone apps, SMSs, and wearable devices. All studies adhered to the standards of randomized controlled trials and used digital devices for objective measurement of physical activity outcome indicators.

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The research findings presented detailed results before, during, and even after follow-up, demonstrating high quality and stability. However, a limitation of this study is the relatively small sample size, which calls for further investigation.

The methodological rigor of the included studies was evaluated using the Cochrane Risk of Bias tool. The results showed that the studies generally had moderate methodological quality, meeting average quality assessment standards. Notably, except for 1 study (Lau et al [46]), all studies explicitly described a robust random sequence generation process for participant allocation, minimizing selection bias. In addition, all studies reported baseline characteristics and used valid, objective measures to assess physical activity outcomes, reducing detection bias.

However, the nature of the interventions posed challenges in achieving blinding of participants and personnel, potentially introducing performance bias. Specifically, although most studies [46,65-67] used intention-to-treat analysis to mitigate attrition bias, the lack of blinding could have influenced participant behavior and data interpretation. Regarding reporting bias, except for Lau et al [46], all studies provided sufficient statistical data for analysis. However, only 5 studies [46,47,50,51,67] reported between-group differences in changes over time, potentially limiting the comprehensiveness of the findings.

Subjective measurement of physical activity and studies involving noncollege student populations were excluded in this analysis, reducing the subjectivity and instability associated with self-report data and enhancing the reliability and consistency of the results. The accuracy of measuring LPA, MVPA, and ST may be influenced by the specific devices and wear-time protocols used in each study. This variability could result in observed inconsistencies. Sensitivity analysis further validated the stability and effectiveness of the meta-analysis results. These rigorous inclusion and exclusion criteria minimized heterogeneity and increased the credibility of the results. However, due to the limited number of included studies, subgroup analysis was not conducted, preventing an understanding of the differences in effects among different populations, intervention measures, and time points.

The key strength of this analysis lies in its focus on the impact of digital health interventions on college students, as well as the inclusion of 4 indicators of physical activity behavior. This may be the first meta-analysis on the effects of digital health interventions specifically targeting college students. However, it is important to note that this study only included English articles, which may have influenced the research sample and compromised the stability of the results. In addition, the included studies mainly originated from Western countries, lacking research on college student populations in non-Western contexts, thus potentially limiting the generalizability of the results. Further research is needed to explore the impact of digital health interventions on physical activity among college students in the future.

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Conclusions

According to the research findings, digital health interventions have shown significant effects on step counts among college students. However, the impact on LPA, moderate to high-intensity physical activity, and sedentary behavior may not be evident. It is important to note that these conclusions are based on a limited number of studies, and further research is needed to validate these findings. Specifically, more research is needed to investigate the impact of digital health interventions on increasing step counts among college students. Therefore, future studies should further explore the effects of digital health interventions on changes in physical activity behavior among college students, particularly in promoting increases in step counts. This will contribute to a better understanding of the effectiveness of digital health interventions and provide scientific evidence for the development of effective intervention strategies.

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Authors' Contributions

SB executed the search strategies. SB and YW conducted the initial screening and full-text review. SB and JY conducted data extraction, while ZWX and ZLQ resolved the conflict. SB conducted narrative analysis and meta-analyses, which were later confirmed by ZYJ and ZR. LI supervised the entire process of the project. All authors contributed to the manuscript drafting.

Conflicts of Interest

None declared.

Multimedia Appendix 1

Searching Strategy. [PDF File (Adobe PDF File), 107 KB-Multimedia Appendix 1]

Multimedia Appendix 2

Study Characteristics. [PDF File (Adobe PDF File), 2661 KB-Multimedia Appendix 2]

Multimedia Appendix 3

PRISMA checklist. [DOCX File , 28 KB-Multimedia Appendix 3]

References

- 1. Arnett JJ. Emerging adulthood. A theory of development from the late teens through the twenties. Am Psychol. 2000;55(5):469-480. [Medline: 10842426]
- Yahia N, Wang D, Rapley M, Dey R. Assessment of weight status, dietary habits and beliefs, physical activity, and nutritional knowledge among university students. Perspect Public Health. 2016;136(4):231-244. [doi: 10.1177/1757913915609945] [Medline: 26475773]
- Aceijas C, Waldhäusl S, Lambert N, Cassar S, Bello-Corassa R. Determinants of health-related lifestyles among university students. Perspect Public Health. 2017;137(4):227-236. [doi: 10.1177/1757913916666875] [Medline: 27597797]
- 4. Park MJ, Scott JT, Adams SH, Brindis CD, Irwin CE. Adolescent and young adult health in the United States in the past decade: little improvement and young adults remain worse off than adolescents. J Adolesc Health. 2014;55(1):3-16. [doi: 10.1016/j.jadohealth.2014.04.003] [Medline: 24815958]
- 5. Piercy KL, Troiano RP, Ballard RM, Carlson SA, Fulton JE, Galuska DA, et al. The physical activity guidelines for Americans. JAMA. 2018;320(19):2020-2028. [FREE Full text] [doi: 10.1001/jama.2018.14854] [Medline: 30418471]
- Abu-Moghli FA, Khalaf IA, Barghoti FF. The influence of a health education programme on healthy lifestyles and practices among university students. Int J Nurs Pract. 2010;16(1):35-42. [doi: <u>10.1111/j.1440-172X.2009.01801.x</u>] [Medline: <u>20158546</u>]
- Wisloff U, Lavie CJ. Taking physical activity, exercise, and fitness to a higher level. Prog Cardiovasc Dis. 2017;60(1):1-2. [doi: <u>10.1016/j.pcad.2017.06.002</u>] [Medline: <u>28793972</u>]
- Lavie CJ, Arena R, Swift DL, Johannsen NM, Sui X, Lee DC, et al. Exercise and the cardiovascular system: clinical science and cardiovascular outcomes. Circ Res. 2015;117(2):207-219. [FREE Full text] [doi: 10.1161/CIRCRESAHA.117.305205] [Medline: 26139859]
- 9. Lavie CJ, Ozemek C, Carbone S, Katzmarzyk PT, Blair SN. Sedentary behavior, exercise, and cardiovascular health. Circ Res. 2019;124(5):799-815. [doi: 10.1161/CIRCRESAHA.118.312669] [Medline: 30817262]

- 10. Global Strategy on Diet, Physical Activity and Health : A Framework to Monitor and Evaluate Implementation. Geneva, Switzerland. World Health Organization; 2004.
- More Physical Activity (RUN). Global action plan on physical activity 2018–2030: more active people for a healthier world. World Health Organization; 2018. URL: <u>https://www.who.int/publications/i/item/9789241514187</u> [accessed 2023-09-19]
- 12. Vuori I. World health organization and physical activity. Prog prev med. 2018;3(1):e0012. [FREE Full text]
- 13. Physical activity. World Health Organization; 2022. URL: <u>https://www.who.int/news-room/fact-sheets/detail/physical-activity</u> [accessed 2021-09-19]
- Donnelly JE, Hillman CH, Castelli D, Etnier JL, Lee S, Tomporowski P, et al. Physical activity, fitness, cognitive function, and academic achievement in children: a systematic review. Med Sci Sports Exerc. 2016;48(6):1197-1222. [FREE Full text] [doi: 10.1249/MSS.0000000000000000] [Medline: 27182986]
- 15. Stubbs B, Vancampfort D, Hallgren M, Firth J, Veronese N, Solmi M, et al. EPA guidance on physical activity as a treatment for severe mental illness: a meta-review of the evidence and position statement from the European psychiatric association (EPA), supported by the international organization of physical therapists in mental health (IOPTMH). Eur Psychiatry. 2018;54:124-144. [doi: 10.1016/j.eurpsy.2018.07.004] [Medline: <u>30257806</u>]
- Ferreira Silva RM, Mendonça CR, Azevedo VD, Raoof Memon A, Noll PRES, Noll M. Barriers to high school and university students' physical activity: a systematic review. PLoS One. 2022;17(4):e0265913. [FREE Full text] [doi: 10.1371/journal.pone.0265913] [Medline: 35377905]
- Iribarren SJ, Cato K, Falzon L, Stone PW. What is the economic evidence for mHealth? A systematic review of economic evaluations of mHealth solutions. PLoS One. 2017;12(2):e0170581. [FREE Full text] [doi: 10.1371/journal.pone.0170581] [Medline: 28152012]
- Hardeman W, Houghton J, Lane K, Jones A, Naughton F. A systematic review of just-in-time adaptive interventions (JITAIs) to promote physical activity. Int J Behav Nutr Phys Act. 2019;16(1):31. [FREE Full text] [doi: 10.1186/s12966-019-0792-7] [Medline: 30943983]
- 19. Monitoring and evaluating digital health interventions: a practical guide to conducting research and assessment. Geneva, Switzerland. World Health Organization; 2016.
- 20. Smith AJ, Skow Á, Bodurtha J, Kinra S. Health information technology in screening and treatment of child obesity: a systematic review. Pediatrics. 2013;131(3):e894-e902. [doi: 10.1542/peds.2012-2011] [Medline: 23382447]
- Montagni I, Cariou T, Feuillet T, Langlois E, Tzourio C. Exploring digital health use and opinions of university students: field survey study. JMIR Mhealth Uhealth. 2018;6(3):e65. [FREE Full text] [doi: 10.2196/mhealth.9131] [Medline: 29549071]
- 22. Tong HL, Coiera E, Laranjo L. Using a mobile social networking app to promote physical activity: a qualitative study of users' perspectives. J Med Internet Res. 2018;20(12):e11439. [FREE Full text] [doi: 10.2196/11439] [Medline: 30578201]
- Müssener U, Thomas K, Bendtsen P, Bendtsen M. High school students' preferences and design recommendations for a mobile phone-based intervention to improve psychological well-being: mixed methods study. JMIR Pediatr Parent. 2020;3(2):e17044. [FREE Full text] [doi: 10.2196/17044] [Medline: 32673268]
- 24. Recommendations on Digital Interventions for Health System Strengthening. Geneva, Switzerland. World Health Organization; 2019.
- 25. WHO Global Observatory for eHealth. mHealth: new horizons for health through mobile technologies: second global survey on eHealth. Geneva, Switzerland. World Health Organization; 2011.
- 26. Berman AH, Gajecki M, Sinadinovic K, Andersson C. Mobile interventions targeting risky drinking among university students: a review. Curr Addict Rep. 2016;3:166-174. [FREE Full text] [doi: 10.1007/s40429-016-0099-6] [Medline: 27226948]
- 27. Dute DJ, Bemelmans WJE, Breda J. Using mobile apps to promote a healthy lifestyle among adolescents and students: a review of the theoretical basis and lessons learned. JMIR Mhealth Uhealth. 2016;4(2):e39. [FREE Full text] [doi: 10.2196/mhealth.3559] [Medline: 27150850]
- 28. Prosser T, Gee KA, Jones F. A meta-analysis of effectiveness of E-interventions to reduce alcohol consumption in college and university students. J Am Coll Health. 2018;66(4):292-301. [doi: 10.1080/07448481.2018.1440579] [Medline: 29452058]
- 29. Åsberg K, Lundgren O, Henriksson H, Henriksson P, Bendtsen P, Löf M, et al. Multiple lifestyle behaviour mHealth intervention targeting Swedish college and university students: protocol for the buddy randomised factorial trial. BMJ Open. 2021;11(12):e051044. [doi: 10.1136/bmjopen-2021-051044]
- Farrer L, Gulliver A, Chan JK, Batterham PJ, Reynolds J, Calear A, et al. Technology-based interventions for mental health in tertiary students: systematic review. J Med Internet Res. 2013;15(5):e101. [FREE Full text] [doi: 10.2196/jmir.2639] [Medline: 23711740]
- Davies EB, Morriss R, Glazebrook C. Computer-delivered and web-based interventions to improve depression, anxiety, and psychological well-being of university students: a systematic review and meta-analysis. J Med Internet Res. 2014;16(5):e130. [FREE Full text] [doi: 10.2196/jmir.3142] [Medline: 24836465]
- 32. Harrer M, Adam SH, Baumeister H, Cuijpers P, Karyotaki E, Auerbach RP, et al. Internet interventions for mental health in university students: a systematic review and meta-analysis. Int J Methods Psychiatr Res. 2019;28(2):e1759. [FREE Full text] [doi: 10.1002/mpr.1759] [Medline: 30585363]

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- Lattie EG, Adkins EC, Winquist N, Stiles-Shields C, Wafford QE, Graham AK. Digital mental health interventions for depression, anxiety, and enhancement of psychological well-being among college students: systematic review. J Med Internet Res. 2019;21(7):e12869. [FREE Full text] [doi: 10.2196/12869] [Medline: 31333198]
- 34. Prensky M. Digital natives, digital immigrants part 1. Horizon J Hum Soc Sci Res. 2001;9(5):1-6. [doi: 10.1108/10748120110424816]
- 35. Prensky M. Digital natives, digital immigrants part 2: do they really think differently? On the Horizon. 2001;9(6):1-6. [doi: 10.1108/10748120110424843]
- 36. Coyne SM, Padilla-Walker LM, Howard E. Emerging in a digital world: a decade review of media use, effects, and gratifications in emerging adulthood. Emerging Adulthood. 2013;1(2):125-137. [doi: 10.1177/2167696813479782]
- 37. Head KJ, Noar SM, Iannarino NT, Grant Harrington N. Efficacy of text messaging-based interventions for health promotion: a meta-analysis. Soc Sci Med. 2013;97:41-48. [doi: 10.1016/j.socscimed.2013.08.003] [Medline: 24161087]
- 38. O'Brien LM. Integrating text-messaging and web-based interventions to improve physical activity and nutrition among college students. Boston, MA. Boston University Libraries; 2014.
- Romeo A, Edney S, Plotnikoff R, Curtis R, Ryan J, Sanders I, et al. Can smartphone apps increase physical activity? systematic review and meta-analysis. J Med Internet Res. 2019;21(3):e12053. [FREE Full text] [doi: 10.2196/12053] [Medline: 30888321]
- 40. Stephens J, Allen J. Mobile phone interventions to increase physical activity and reduce weight: a systematic review. J Cardiovasc Nurs. 2013;28(4):320-329. [FREE Full text] [doi: 10.1097/JCN.0b013e318250a3e7] [Medline: 22635061]
- 41. Müller AM, Alley S, Schoeppe S, Vandelanotte C. Int J Behav Nutr Phys Act. 2016;13(1):109. [FREE Full text] [doi: 10.1186/s12966-016-0434-2] [Medline: 27724911]
- 42. Gal R, May AM, van Overmeeren EJ, Simons M, Monninkhof EM. The effect of physical activity interventions comprising wearables and smartphone applications on physical activity: a systematic review and meta-analysis. Sports Med Open. 2018;4(1):42. [FREE Full text] [doi: 10.1186/s40798-018-0157-9] [Medline: 30178072]
- 43. Kim HN, Seo K. Smartphone-based health program for improving physical activity and tackling obesity for young adults: a systematic review and meta-analysis. Int J Environ Res Public Health. 2019;17(1):15. [FREE Full text] [doi: 10.3390/ijerph17010015] [Medline: 31861359]
- 44. Wibowo RA, Kelly P, Baker G. The effect of smartphone application interventions on physical activity level among university/college students: a systematic review protocol. Physical Therapy Reviews. 2020;25(2):135-142. [doi: 10.1080/10833196.2020.1756125]
- 45. Hebden L, Cook A, van der Ploeg HP, King L, Bauman A, Allman-Farinelli M. A mobile health intervention for weight management among young adults: a pilot randomised controlled trial. J Hum Nutr Diet. 2014;27(4):322-332. [doi: 10.1111/jhn.12155] [Medline: 23992038]
- 46. Lau PWC, Wang JJ, Ransdell LL, Shi L. The effectiveness of facebook as a social network intervention to increase physical activity in Chinese young adults. Front Public Health. 2022;10:912327. [FREE Full text] [doi: 10.3389/fpubh.2022.912327] [Medline: 35937270]
- 47. Kim Y, Lumpkin A, Lochbaum M, Stegemeier S, Kitten K. Promoting physical activity using a wearable activity tracker in college students: a cluster randomized controlled trial. J Sports Sci. 2018;36(16):1889-1896. [doi: 10.1080/02640414.2018.1423886] [Medline: 29318916]
- Davies CA, Spence JC, Vandelanotte C, Caperchione CM, Mummery W. Meta-analysis of internet-delivered interventions to increase physical activity levels. Int J Behav Nutr Phys Act. 2012;9:52. [FREE Full text] [doi: 10.1186/1479-5868-9-52] [Medline: 22546283]
- 49. Beleigoli AM, Andrade AQ, Cançado AG, Paulo MN, Diniz MDFH, Ribeiro AL. Web-based digital health interventions for weight loss and lifestyle habit changes in overweight and obese adults: systematic review and meta-analysis. J Med Internet Res. 2019;21(1):e298. [FREE Full text] [doi: 10.2196/jmir.9609] [Medline: 30622090]
- 50. Pope ZC, Gao Z. Feasibility of smartphone application- and social media-based intervention on college students' health outcomes: a pilot randomized trial. J Am Coll Health. 2022;70(1):89-98. [doi: 10.1080/07448481.2020.1726925] [Medline: 32150514]
- Pope ZC, Barr-Anderson DJ, Lewis BA, Pereira MA, Gao Z. Use of wearable technology and social media to improve physical activity and dietary behaviors among college students: a 12-week randomized pilot study. Int J Environ Res Public Health. 2019;16(19):3579. [FREE Full text] [doi: 10.3390/ijerph16193579] [Medline: 31557812]
- 52. Mclaughlin M, Delaney T, Hall A, Byaruhanga J, Mackie P, Grady A, et al. Associations between digital health intervention engagement, physical activity, and sedentary behavior: systematic review and meta-analysis. J Med Internet Res. 2021;23(2):e23180. [FREE Full text] [doi: 10.2196/23180] [Medline: 33605897]
- 53. Strain T, Wijndaele K, Dempsey PC, Sharp SJ, Pearce M, Jeon J, et al. Wearable-device-measured physical activity and future health risk. Nat Med. 2020;26(9):1385-1391. [FREE Full text] [doi: 10.1038/s41591-020-1012-3] [Medline: 32807930]
- Goodyear VA, Wood G, Skinner B, Thompson JL. The effect of social media interventions on physical activity and dietary behaviours in young people and adults: a systematic review. Int J Behav Nutr Phys Act. 2021;18(1):72. [FREE Full text] [doi: 10.1186/s12966-021-01138-3] [Medline: 34090469]

https://www.jmir.org/2024/1/e51714

RenderX

- Page MJ, McKenzie JE, Bossuyt PM, Boutron I, Hoffmann TC, Mulrow CD, et al. The PRISMA 2020 statement: an updated guideline for reporting systematic reviews. BMJ. 2021;372:n71. [FREE Full text] [doi: 10.1136/bmj.n71] [Medline: 33782057]
- 56. Liberati A, Altman DG, Tetzlaff J, Mulrow C, Gøtzsche PC, Ioannidis JPA, et al. The PRISMA statement for reporting systematic reviews and meta-analyses of studies that evaluate healthcare interventions: explanation and elaboration. BMJ. 2009;339:b2700. [FREE Full text] [doi: 10.1136/bmj.b2700] [Medline: 19622552]
- 57. Higgins JPT, Deeks JJ. Selecting studies and collecting data. In: Cochrane Handbook for Systematic Reviews of Interventions: Cochrane Book Series. Hoboken, New Jersey. John Wiley & Sons, Ltd; 2008:151-185.
- Cumpston M, Li T, Page MJ, Chandler J, Welch VA, Higgins J, et al. Updated guidance for trusted systematic reviews: a new edition of the cochrane handbook for systematic reviews of interventions. Cochrane Database Syst Rev. 2019;10(10):ED000142. [FREE Full text] [doi: 10.1002/14651858.ED000142] [Medline: 31643080]
- Higgins JPT, Altman DG, Gøtzsche PC, Jüni P, Moher D, Oxman AD, Cochrane Bias Methods Group, et al. Cochrane Statistical Methods Group. The cochrane collaboration's tool for assessing risk of bias in randomised trials. BMJ. 2011;343:d5928. [FREE Full text] [doi: 10.1136/bmj.d5928] [Medline: 22008217]
- 60. Higgins JPT, Thompson SG. Quantifying heterogeneity in a meta-analysis. Stat Med. 2002;21(11):1539-1558. [doi: 10.1002/sim.1186] [Medline: 12111919]
- 61. Schulz KF, Altman DG, Moher D, CONSORT Group. CONSORT 2010 statement: updated guidelines for reporting parallel group randomised trials. BMJ. 2010;340:c332. [FREE Full text] [doi: 10.1136/bmj.c332] [Medline: 20332509]
- 62. Higgins JPT, Thompson SG, Deeks JJ, Altman DG. Measuring inconsistency in meta-analyses. BMJ. 2003;327(7414):557-560. [FREE Full text] [doi: 10.1136/bmj.327.7414.557] [Medline: 12958120]
- 63. Lau J, Ioannidis JP, Schmid CH. Quantitative synthesis in systematic reviews. Ann Intern Med. 1997;127(9):820-826. [doi: 10.7326/0003-4819-127-9-199711010-00008] [Medline: 9382404]
- 64. Cohen J. Statistical Power Analysis for the Behavioral Sciences 2nd ed. New York. Routledge; 1988.
- Kellner M, Dold C, Lohkamp M. Objectively assessing the effect of a messenger-based intervention to reduce sedentary behavior in university students: a pilot study. J Prev (2022). 2023;44(5):521-534. [FREE Full text] [doi: 10.1007/s10935-023-00735-1] [Medline: 37171555]
- 66. Miragall M, Domínguez-Rodríguez A, Navarro J, Cebolla A, Baños RM. Increasing physical activity through an internet-based motivational intervention supported by pedometers in a sample of sedentary students: a randomised controlled trial. Psychol Health. 2018;33(4):465-482. [doi: 10.1080/08870446.2017.1368511] [Medline: 28880576]
- 67. Al-Nawaiseh HK, McIntosh WA, McKyer LJ. An-m-health intervention using smartphone app to improve physical activity in college students: a randomized controlled trial. Int J Environ Res Public Health. 2022;19(12):7228. [doi: 10.3390/ijerph19127228] [Medline: 35742477]
- 68. Sasaki JE, John D, Freedson PS. Validation and comparison of ActiGraph activity monitors. J Sci Med Sport. 2011;14(5):411-416. [doi: 10.1016/j.jsams.2011.04.003] [Medline: 21616714]
- Kozey SL, Staudenmayer JW, Troiano RP, Freedson PS. Comparison of the ActiGraph 7164 and the ActiGraph GT1M during self-paced locomotion. Med Sci Sports Exerc. 2010;42(5):971-976. [FREE Full text] [doi: 10.1249/MSS.0b013e3181c29e90] [Medline: 19997000]
- Jakicic JM, Marcus M, Gallagher KI, Randall C, Thomas E, Goss FL, et al. Evaluation of the SenseWear pro armband to assess energy expenditure during exercise. Med Sci Sports Exerc. 2004;36(5):897-904. [doi: 10.1249/01.mss.0000126805.32659.43] [Medline: 15126727]
- Lee KY, Macfarlane DJ, Cerin E. Comparison of three models of actigraph accelerometers during free living and controlled laboratory conditions. Eur J Sport Sci. 2013;13(3):332-339. [FREE Full text] [doi: 10.1080/17461391.2011.643925] [Medline: 23679150]
- 72. Migueles JH, Cadenas-Sanchez C, Ekelund U, Delisle Nyström C, Mora-Gonzalez J, Löf M, et al. Accelerometer data collection and processing criteria to assess physical activity and other outcomes: a systematic review and practical considerations. Sports Med. 2017;47(9):1821-1845. [FREE Full text] [doi: 10.1007/s40279-017-0716-0] [Medline: 28303543]
- 73. O'Brien MW, Wu Y, Petterson JL, Bray NW, Kimmerly DS. Validity of the ActivPAL monitor to distinguish postures: a systematic review. Gait Posture. 2022;94:107-113. [doi: 10.1016/j.gaitpost.2022.03.002] [Medline: 35276456]
- 74. Paluch AE, Bajpai S, Bassett DR, Carnethon MR, Ekelund U, Evenson KR, et al. Steps for Health Collaborative. Daily steps and all-cause mortality: a meta-analysis of 15 international cohorts. Lancet Public Health. 2022;7(3):e219-e228. [FREE Full text] [doi: 10.1016/S2468-2667(21)00302-9] [Medline: 35247352]
- 75. Bravata DM, Smith-Spangler C, Sundaram V, Gienger AL, Lin N, Lewis R, et al. Using pedometers to increase physical activity and improve health: a systematic review. JAMA. 2007;298(19):2296-2304. [doi: <u>10.1001/jama.298.19.2296</u>] [Medline: <u>18029834</u>]
- 76. Warburton DER, Nicol CW, Bredin SSD. Health benefits of physical activity: the evidence. CMAJ. 2006;174(6):801-809. [FREE Full text] [doi: 10.1503/cmaj.051351] [Medline: 16534088]
- 77. Tudor-Locke C, Bassett DR. How many steps/day are enough? Preliminary pedometer indices for public health. Sports Med. 2004;34(1):1-8. [doi: 10.2165/00007256-200434010-00001] [Medline: 14715035]

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- Murtagh EM, Murphy MH, Boone-Heinonen J. Walking: the first steps in cardiovascular disease prevention. Curr Opin Cardiol. 2010;25(5):490-496. [FREE Full text] [doi: 10.1097/HCO.0b013e32833ce972] [Medline: 20625280]
- 79. Moore CC, Evenson KR, Shiroma EJ, Howard AG, Cuthbertson CC, Buring JE, et al. Abstract 069: sporadic step accumulation and all-cause mortality in the women's health study: do all steps count? Circulation. 2021;143(Suppl_1):A069.
- 80. Warburton DER, Bredin SSD. Reflections on physical activity and health: what should we recommend? Can J Cardiol. 2016;32(4):495-504. [doi: <u>10.1016/j.cjca.2016.01.024</u>] [Medline: <u>26995692</u>]
- 81. Warburton DER, Bredin SSD. Health benefits of physical activity: a systematic review of current systematic reviews. Curr Opin Cardiol. 2017;32(5):541-556. [doi: 10.1097/HCO.00000000000437] [Medline: 28708630]
- 82. Number of steps per day more important than step intensity. National Institutes of Health URL: <u>https://www.nih.gov/news-events/nih-research-matters/number-steps-day-more-important-step-intensity</u> [accessed 2023-09-19]
- 83. Total daily steps, not step intensity, offer more benefits. Harvard Health Publishing; 2020. URL: <u>https://www.</u> health.harvard.edu/staying-healthy/total-daily-steps-not-step-intensity-offer-more-benefits [accessed 2023-09-19]
- Buman MP, Hekler EB, Haskell WL, Pruitt L, Conway TL, Cain KL, et al. Objective light-intensity physical activity associations with rated health in older adults. Am J Epidemiol. 2010;172(10):1155-1165. [FREE Full text] [doi: 10.1093/aje/kwq249] [Medline: 20843864]
- 85. Chastin SFM, Palarea-Albaladejo J, Dontje ML, Skelton DA. Combined effects of time spent in physical activity, sedentary behaviors and sleep on obesity and cardio-metabolic health markers: a novel compositional data analysis approach. PLoS One. 2015;10(10):e0139984. [FREE Full text] [doi: 10.1371/journal.pone.0139984] [Medline: 26461112]
- 86. Chastin SFM, De Craemer M, De Cocker K, Powell L, Van Cauwenberg J, Dall P, et al. How does light-intensity physical activity associate with adult cardiometabolic health and mortality? Systematic review with meta-analysis of experimental and observational studies. Br J Sports Med. 2019;53(6):370-376. [FREE Full text] [doi: 10.1136/bjsports-2017-097563] [Medline: 29695511]
- Farrahi V, Rostami M, Dumuid D, Chastin SFM, Niemelä M, Korpelainen R, et al. Joint profiles of sedentary time and physical activity in adults and their associations with cardiometabolic health. Med Sci Sports Exerc. 2022;54(12):2118-2128. [FREE Full text] [doi: 10.1249/MSS.000000000003008] [Medline: 35881930]
- Loprinzi PD, Cardinal BJ. Measuring children's physical activity and sedentary behaviors. J exerc sci fit. 2011;9(1):15-23. [doi: <u>10.1016/s1728-869x(11)60002-6</u>]
- Tudor-Locke C, Craig CL, Brown WJ, Clemes SA, De Cocker K, Giles-Corti B, et al. How many steps/day are enough? For adults. Int J Behav Nutr Phys Act. 2011;8:79. [FREE Full text] [doi: 10.1186/1479-5868-8-79] [Medline: 21798015]

Abbreviations

CONSORT: Consolidated Standards of Reporting Trials eHealth: electronic health LPA: light intensity physical activity mHealth: mobile health MVPA: moderate to vigorous physical activity PRISMA: Preferred Reporting Items for Systematic Reviews and Meta-Analyses PROSPERO: International Prospective Register of Systematic Reviews SMD: standardized mean difference ST: sedentary time

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