

# Can mHealth Revolutionize the Way We Manage Adult Obesity?

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## Abstract

Obesity is the largest driver of chronic preventable diseases, accounting for an estimated \$147 billion or 10 percent of total US healthcare costs in 2008. It has been forecasted that 42 percent of Americans will be obese by 2030. Mobile health (mHealth) technologies target and may modify the behavioral factors that lead to obesity to promote a healthy lifestyle. These technologies could potentially reduce the cost and the morbidity and mortality burden of obesity because of their inexpensive and portable nature. This study aimed to analyze the efficacy and cost-effectiveness of mHealth interventions for adult obesity in the United States. The methodology used in this study was a literature review of 54 articles. Weight, body mass index (BMI), waist circumference reductions, and favorable lifestyle behavior changes were noted across most studies. Existing data and research on efficacy and linked costs indicated that mHealth technologies were more effective than other methods and could be inexpensively delivered remotely to manage adult obesity, offering significant benefits over conventional care. Further studies on the costs and benefits of adapting such mHealth interventions in clinical settings are needed.

**Keywords:** adult, obesity, mHealth interventions, cost, efficacy

## Introduction

Obesity—declared an epidemic in 1999—is the single greatest public health threat in the United States.<sup>1</sup> In 2015, adult obesity rates exceeded 35 percent four states—Alabama, Louisiana, Mississippi, and West Virginia. Moreover, the rates were above 20 percent in other states, including: Georgia, Kentucky, Illinois, Oregon, Pennsylvania, and Texas.<sup>2</sup> If current trends continue, the national obesity rate has been projected to climb from 35.7 percent in 2012 to 42 percent by 2030.<sup>3</sup>

Obesity increases the risk of serious comorbidities such as diabetes mellitus, cardiovascular diseases, and endometrial cancer, thereby resulting in high morbidity, mortality, and healthcare costs.<sup>4-6</sup> In 2008, these obesity-linked healthcare costs were an estimated \$147 billion, constituting 10 percent of the national medical budget.<sup>7</sup> In 2011, a study using a simulation model predicted that healthcare expenses due to obesity-related preventable diseases in the United States will rise by \$48–66 billion per year by 2030.<sup>8</sup>

Lifestyle modifications consisting of behavioral interventions such as diet monitoring, exercise program, and counseling have resulted in clinically significant weight loss in obese people.<sup>9</sup> However, long-term adherence to these modifications has often been quite expensive, time consuming, and challenging.<sup>10</sup>

Mobile health (mHealth) has been defined by the World Health Organization as medical and public health practice supported by mobile devices, such as mobile phones, patient monitoring devices, personal digital assistants, and other wireless devices.<sup>11</sup> It can be categorized into six categories: education, clinical decision support, remote data collection and analysis, health promotion and awareness, remote monitoring, and integrated care and diagnostic support.<sup>12</sup>

Utilization of mHealth has surged because of its portable, accessible, and cost-effective nature.<sup>13-15</sup> For instance, short message services (SMS) is a popular, low-cost, and simple way to deliver health information.<sup>16</sup> Likewise, smartphones have gained popularity and are being adopted for the prevention and control of obesity because they offer multiple functionalities.<sup>17</sup> Different types of mobile applications have already been developed for general use in obesity management, utilizing features such as movement sensors, microphones and cameras.<sup>18,19</sup> For example, smartphone software could utilize motion sensors

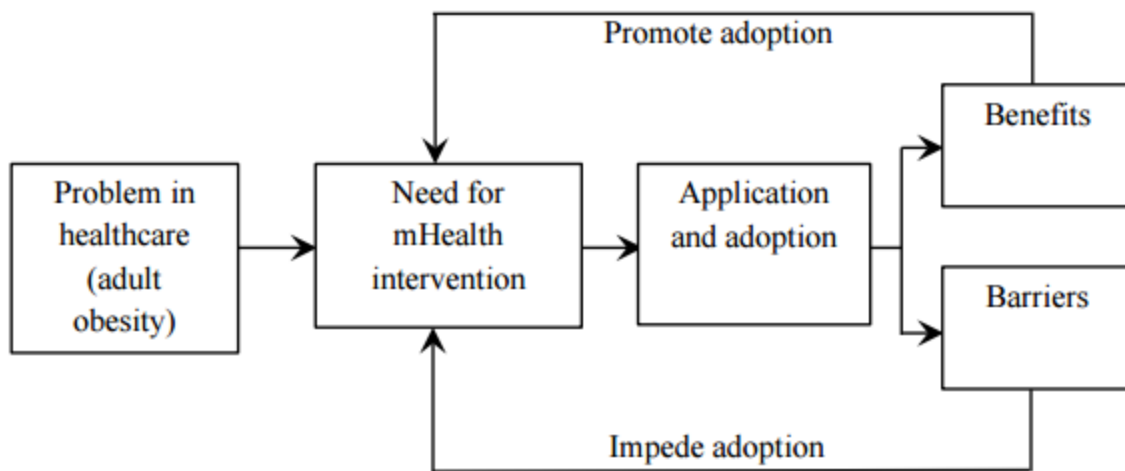
and GPS to create maps of exercise routes and provide users with real-time feedback regarding movement speed, step counts, energy expenditure, and the completion of exercise goals. In addition, diet cameras could recognize foods and calculate the calorie content of a meal automatically from images or videos, requiring limited human input.<sup>20,21</sup>

Thus, mobile technology has proffered an exciting opportunity to remotely deliver obesity interventions.<sup>22</sup> This opportunity raises the question as to whether such interventions are truly better—more effective and cheaper—compared with conventional practices, making it crucial to assess their effectiveness and costs. Therefore, the purpose of this study was to identify mobile interventions currently directed toward adult obesity in the United States and to evaluate their efficacy and related costs.

## Methodology

The conceptual framework for this research conformed to the steps and research framework used by Yao, Chu, and Li (2010).<sup>23</sup> The framework elucidates the course of mHealth adoption for management of the adult obesity epidemic. If the process of mHealth programs' leading to improved care access for obese adults while diminishing parallel healthcare expenses is to be investigated, their effectiveness and corresponding costs must first be pinpointed. Akin to the cyclic progression of any project, the process of technology adoption begins when problems in the existing system necessitate assessment of needs, which subsequently results in the creation and institution of a solution. The solution here is the application of mHealth interventions. The process involves an evaluation of the benefits of and barriers to mHealth utilization after it is adopted, and the evaluation is repeated to assess the benefits and barriers related to the technology (see [Figure 1](#)). This conceptual framework is suitable for the present study because it centers on means of application of new technology in healthcare settings. Moreover, the internal validity of this approach is supported by its successful replication in past studies.<sup>24-26</sup>

**Figure 1: Research Framework**



Source: Adapted from Yao, W., C. H. Chu, and Z. Li. "The Use of RFID in Healthcare: Benefits and Barriers." *Proceedings of the 2010 IEEE International Conference on RFID Technology and Applications (RFID-TA)* (2010): 128–34.

The hypothesis of this study was that mHealth interventions would demonstrate greater efficacy with minimal expense in preventing and controlling obesity when compared with standard care.

A systematic search methodology was used in this literature review. For the intent of this research question, a comprehensive and exhaustive systematic review or meta-analysis was not feasible because of the abundance of studies of heterogeneous quality. The literature review was conducted in three distinct stages:

1. determining the search strategy, and identifying and collecting literature;
2. establishing inclusion criteria, scrutinizing text for relevancy, and analyzing the literature data; and
3. identifying appropriate categories.

### Step 1: Literature Identification and Collection

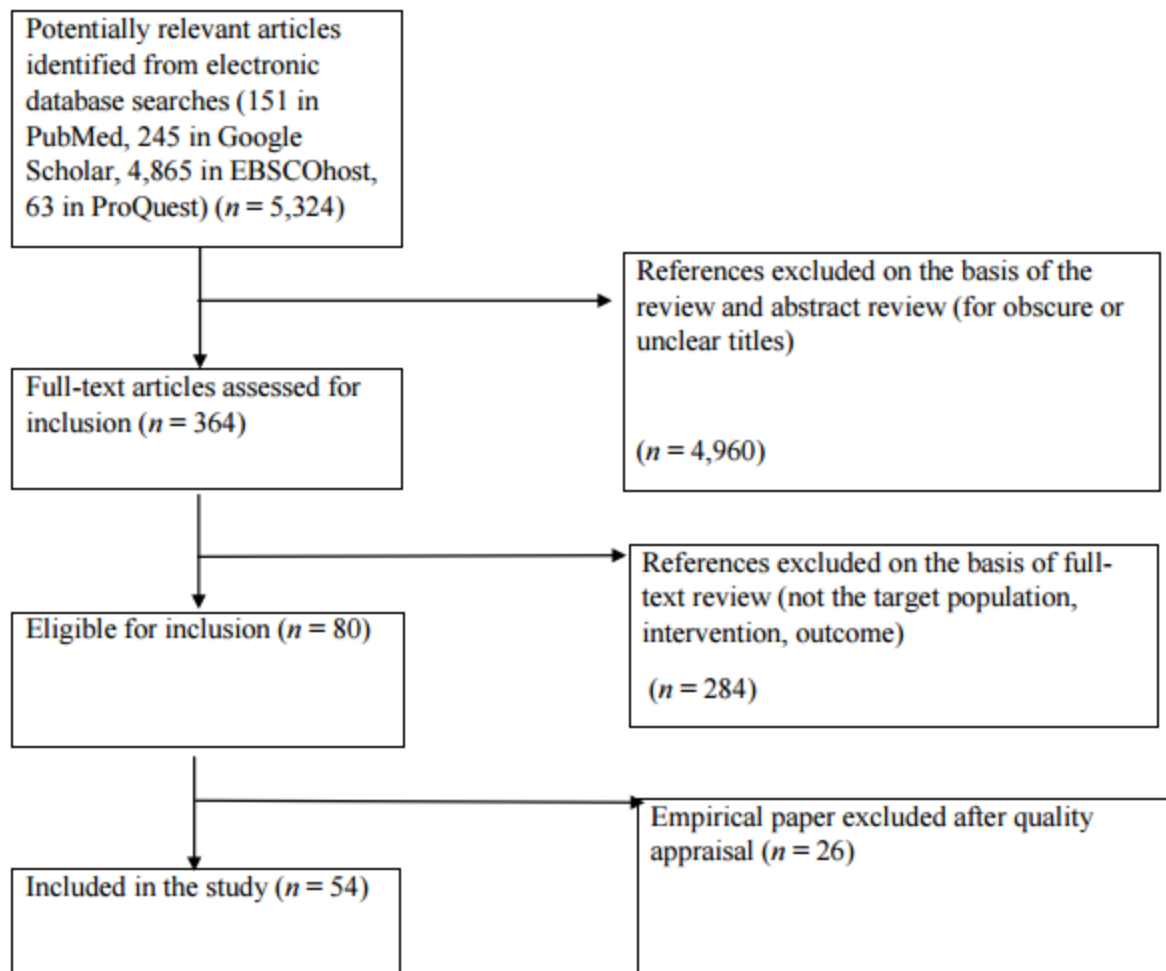
The electronic databases PubMed, EBSCOhost, ProQuest, Academic Search Premier, and Google Scholar were searched for the following terms: “mobile health interventions” or “mobile health apps” or “smart phone apps” and “obesity” and “Adults and “weight loss” or “physical activity” or “exercise” or “sedentary behavior” and “efficacy” or “cost-effectiveness” or “economic evaluation.” *The International Journal of Medical Informatics*, *International Journal of Obesity*, *International Journal of Research in Medical Sciences*, and other reliable healthcare websites, including those of the National Institutes of Health and the World Health Organization, were also used. Citations and abstracts identified in the search were also assessed to identify relevant articles.

## Step 2: Establishment of Inclusion Criteria and Literature Analysis

As the utilization of mobile technology within healthcare increases, it is important to assess the effectiveness and value for the cost of the various mobile devices. Consequently, the literature was selected to include the efficacy and cost-effectiveness of the mobile health interventions for obesity. In an attempt to stay current in the research study, the search was limited to sources published since 2006. The search was also limited to sources attainable as full texts and to studies written in the English language and conducted in the United States.

The methodology and results of the identified texts were analyzed, and key papers were identified and included within the research query. References were reviewed and determined to have satisfied the inclusion criteria if the material provided accurate information about physical activity, BMI, waist circumference, lifestyle behavior, and cost-effectiveness. From a total of 80 references found, 54 references were selected for this research study. A summary of the selection criteria and process is displayed in [Figure 2](#). The literature search was conducted by B.W., K.G., and N.B. and was validated by A.C., who acted as a second reader and also double-checked that the references met the research study’s inclusion criteria. D.P. reviewed the manuscript and revised it for clarity and consistency.

**Figure 2: Flow Diagram for Selection of Studies on mHealth Interventions for Adult Obesity in the United States, 2006-1016**



### Step 3: Literature Categorization

The results were sorted into the following categories: study characteristics, effect on body weight, effect on waist circumference and BMI, effect on lifestyle behavior (change in dietary intake and change in physical activity), effect on adherence and satisfaction, and cost efficacy.

## Results

A summary of studies that examined the effects of mobile apps on various aspects of obesity appears below. [Table 1](#) describes characteristics of the studies that were examined, [Table 2](#) lists the distribution of participants by gender and race for included studies, and [Table 3](#) summarizes the effects of each study on the output parameters of the studies. The mHealth interventions included in the studies broadly utilized the following avenues to deliver the intervention: health information delivery, such as education or motivation ( $n = 17$ ); reminders ( $n = 6$ ); communication platform for patients to communicate with healthcare providers ( $n = 3$ ); self-monitoring, such as patients' recording of their diet or exercise trends ( $n = 12$ ); peer or group support ( $n = 7$ ); and remote monitoring, in which patient data are collected through remote peripheral monitoring devices ( $n = 10$ ). Some studies were multifaceted and involved more than one of these types of intervention.

### Effect on Body Weight

Because body weight is the most important parameter associated with obesity, it is reasonable that a majority of the studies examined this parameter directly. A total of 9 of the 14 mobile intervention studies that reported body weight changes found that the intervention group lost more weight than the control group. In these studies, the range of weight loss in the intervention group was 1.27 kg to 10.9 kg, compared with -1.14 kg to 2.45 kg for the control group. The Weighing Everyday to Improve and Gain Health plan, for example, used a smartphone-enabled scale for daily weighing, with a web-based weight loss graph and weekly e-mails with tailored feedback and lessons, and a study of the plan found that the intervention group lost more weight compared with the control group at 6 months (6.55 percent [7.7 kg] vs. 0.35 percent [1.5 kg]) and that a greater percentage achieved 5 percent weight loss (42.6 percent of the intervention group vs. 6.8 percent of the control group) and 10 percent weight loss (27.7 percent of the intervention group vs. 0 percent of the control group).<sup>27</sup> The Health-E-Call treatment, which included a smartphone-based component (Health-E-Call app) and a minimal in-person component, resulted in an average weight loss of 10.9 kg, with 90 percent and 70 percent of participants reaching the 5 percent and 10 percent weight-loss milestones at 24 weeks, respectively.<sup>28</sup> In another study, college students in a group that received Facebook plus text messaging and personalized feedback lost 2.4 kg compared with 0.63-kg loss in the Facebook-only group and 0.24-kg loss in the waiting-list groups at eight weeks.<sup>29</sup> In contrast, the Mobile Pounds Off Digitally study—which required the one group to use a diet and physical activity monitoring app on their mobile device and to interact with study counselors and other participants on Twitter in addition to using the podcasts that the podcast-only group used—demonstrated an equal degree of weight loss (2.7 percent each) in the two groups.<sup>30</sup> (See [Table 1](#).)

Out of the four interventions employing other mHealth devices, two had significant findings favoring the intervention. During a study that tested whether the outcome of physician-directed weight-loss treatment could be improved by adding mobile technology, the mobile technology group lost a mean of 3.9 kg more than the group that received the standard treatment.<sup>31</sup> The Self-Monitoring and Recording using Technology trial found mean weight loss of 2.32 percent with use of personal digital assistants (PDAs) with feedback for weight self-monitoring, while the control group lost 1.94 percent.<sup>32</sup> Another study demonstrated weight loss of 1.86 kg in the group weight-loss education group, 3.55 kg in a group that used armbands, and 6.59 kg in the group that used both of these approaches.<sup>33</sup> (See [Table 3](#).)

Evidence found during a systematic review that investigated the benefits of self-regulatory smartphone apps for weight loss in obese adults was suggestive of the usefulness of these apps for diet self-regulation for weight loss because participants in the smartphone application group in all studies lost at least some body weight. However, no significant difference in the weight loss was found in comparison with other self-monitoring methods.<sup>34</sup>

The authors of a study of innovative interventions to foster behavioral change in overweight and obese individuals noted three systematic reviews of the use of mHealth to address obesity. Two of these reviews concluded that mHealth was associated with moderate weight loss and positive behavioral changes, while one review showed that the use of SMS data offers numerous methodological benefits over conventional behavioral therapy despite being in its early stages.<sup>35</sup>

## Effects on Waist Circumference and BMI

Three studies that tracked waist circumference reported reduction in waist circumference, while three of five trials measuring BMI documented a decrease in BMI. The Smart Coach for Lifestyle Management trial noted that waist circumference reductions in the intensive counseling/smartphone group were 7.01 cm in males and 5.68 cm in females, compared with 3.00 cm in males and 3.19 cm in females in the control group. Reductions in BMI of 1.3 kg/m<sup>2</sup> in the intensive counseling/smartphone group and 1.4 kg/m<sup>2</sup> in the control groups were seen.<sup>36</sup> In the Self-Monitoring and Recording using Technology trial, the highest percentage reduction of waist circumference—6.4 percent—occurred in that group that used PDAs with feedback, whereas in the PDA-only and control groups the reductions in waist circumference were 5.0 percent and 4.0 percent, respectively.<sup>37</sup> In another study, the intervention group using a mobile app called YouPlus reduced their mean waist circumference by 7.2 cm, or 6.6 percent.<sup>38</sup> (See [Table 3](#).)

In a post hoc analysis of the effects of an app called Mobile Pounds Off Digitally, app users reported significantly lower BMI at 6 months (31.5 kg/m<sup>2</sup>) than nonusers (32.5 kg/m<sup>2</sup>).<sup>39</sup> Examination of the effect of the Mobile Audio Manager and Response Tracker system app, which utilized Episodic Future Thinking (EFT) as the behavior-modifying technique to induce weight loss in a pilot study, demonstrated a higher decline in BMI (1.0 vs. 0.2 kg/m<sup>2</sup>) in the EFT group versus controls. App users also stated that the system was useful, helpful, and easy to use.<sup>40</sup> (See [Table 3](#).)

## Effects on Lifestyle Behavior: Change in Dietary Behavior

Changes in eating behavior with or without changes in fruit and vegetable intake were examined in six studies.<sup>41-46</sup> In a study of a four-month mDiet intervention wherein the experimental group received two to five weight-management text messages per day and a comparison group received the usual care, with the use of three 24-hour recalls to assess fruit/vegetable intake change and the use of eating behavior inventory (EBI) scores to measure the change in eating behaviors, the authors concluded that moderate short-term weight loss (2.3 kg vs. 0.63 kg) was achieved by sending text messages that promoted healthy eating strategies.<sup>47</sup> In the Mobile Pounds Off Digitally study, users of the app and podcasts had a nonsignificantly higher EBI score of 12.4 points with lower fruit intake of 1.4 cups per day and vegetable intake of 2.2 cups per day, compared with podcast-only users' 9.8 points, 1.6 cups of fruit per day, and 2.6 cups of vegetables per day.<sup>48</sup> In the Smart Coach for Lifestyle Management study, it was found that the number of fruit and vegetable servings per day increased in all groups (0.81 for intensive counseling, 0.51 for intensive counseling/smartphone, and 2.1 for less-intensive counseling/smartphone) except for the smartphone-only group, which had a slight decrease (0.05).<sup>49</sup>

An investigation of the ability of the MyFitnessPal app to influence dietary behavior in primary care patients found that the intervention group receiving app download assistance and 6 months of usual care used a “personal calorie goal” for an extra two days per week compared with the control participants receiving no such assistance.<sup>50</sup> In a study in which a smartphone app intervention tool (SMARTAPPetite) targeted change in eating behavior as a primary outcome, a strong association between program participation and improvement in healthy eating was seen, including a direct effect on the consumption of healthy foods.<sup>51</sup> In another study, the Episodic Future Thinking group found the Mobile Audio Manager and Response Tracker system more helpful than other methods in affecting eating behavior, with Likert scale ratings of 4.3 in that group versus 3.9 in the control group.<sup>52</sup> (See [Table 3](#).)

Daily caloric intake changes were measured in several studies, but none of the results were statistically significant. Decreases in caloric intake of 415.6, 468.2, 218.5, and 249.2 kcal/day in the intensive counseling, intensive counseling/smartphone, less-intensive counseling/smartphone, and smartphone-only groups, respectively, were demonstrated in the Smart Coach for Lifestyle Management trial.<sup>53</sup> Although none of their outcome measures were significantly different among any of the interventions studied, the researchers concluded that the results of their pilot trial of a weight-loss intervention using smartphone technology provided preliminary support for use of a smartphone app for self-monitoring as an adjunct to behavioral counseling.<sup>54</sup> The Mobile Audio Manager and Response Tracker system trial found larger reductions in energy intake of 791.2 kcal/day in the Episodic Future Thinking group versus 482.4 kcal/day in the control group.<sup>55</sup> In the Weighing Every Day to Improve and Gain Health plan, the treatment group consumed fewer calories (1,509 kcal/day) compared with control participants (1,856 kcal/day).<sup>56</sup> In the Mobile Pounds Off Digitally study, the group using podcasts and the app minimized their energy intake by 288.8 kcal/day, with controls demonstrating a decrease of 242.5 kcal/day.<sup>57</sup> (See [Table 3](#).)

## Effects on Lifestyle Behavior: Change in Physical Activity

In five out of nine studies, physical activity levels increased in the mobile device intervention groups relative to the controls. Assessment of physical activity by recording the step counts was done during the Text4Diet trial. Compared to the controls (which received only monthly newsletters), the experimental group receiving daily interactive SMS and MMS increased their step count to 3000 steps per day from their baseline levels, and these higher step counts were associated with greater weight loss. There was a lack of group weight loss differences over 12 months (2.27 lb vs 3.64 lb; control vs intervention) but an improvement in weight-related behaviors and weight outcomes was noted.<sup>58</sup> In a study over a much shorter time period (6 weeks) that evaluated effects of thrice weekly text messaging on older non-Hispanic blacks adults, small improvements of 679 and 398 in daily steps and of 11.96 and 4.55 points in perceived activity levels (leisure time exercise questionnaire scores) were seen in the experimental and control group, respectively.<sup>59</sup>

One study developed and tested three types (social, analytic and affective) of mobile apps in a randomized clinical trial. Participants across all three apps reported significant mean increases of 100.8 minutes/week in brisk walking and of 188.6 minutes/week in MVPA across the eight-week intervention period and significant decreases in the daily amount of discretionary time spent viewing television. All three apps were found to be generally easy to use and acceptable by the participants, who had no prior experience with smartphones.<sup>60</sup> Another study reported an average increase of 100.8 mins/week in brisk walking and 188.6 mins/week in MVPA for its three intervention groups.<sup>61</sup> While daily total step increased by approximately 800 or 15 percent over three weeks in the study whose participants that were encouraged daily via the mobile phone prompts to increase steps by 20 percent from the previous week.<sup>62</sup> Post-hoc analysis in the Mobile Pounds Off Digitally study revealed app users had higher exercise self-monitoring rates (2.6 vs 1.2 days/week) and intentional physical activity (196.4 vs 100.9 kcal/day) than non-app users.<sup>63</sup>

The ability of B-Mobile app, that prompts users to take physical activity, breaks and delivers feedback on Excessive Sedentary Time (SED) which is a surrogate measure for obesity, to decrease SED and increase Low Intensity Physical Activity (LPA) and Moderate to Vigorous Intensity Physical Activity (MVPA) was evaluated in a study. A decrease in the average percentage SED by 5.9 percent, 5.6 percent and 3.3 percent was noted in three-minute, six-minute and 12-minute activity breaks, respectively. Also, an increase in percentage LPA by 3.9 percent, 3.9 percent, 1.9 percent, and percent MVPA by 2 percent, 1.7 percent and 1.3 percent, respectively, was seen in these three groups.<sup>64</sup> A combination of Exergame smartphone apps and motivational texts was evaluated for 12 weeks with results that found the intervention group's activity levels dropped lesser than controls (-356.8 vs -722 min/week MET).<sup>65</sup> (See [Table 3.](#))

## Effect on Adherence and Satisfaction

All ten studies measuring program adherence recorded greater levels in the intervention group compared to the control, with four noting association of greater adherence with greater weight loss. Adherence in the intervention groups ranged from 60 percent-85 percent. On specific examination of impact of three mobile apps on adherence and satisfaction, all participants found each of the apps easy to use and somewhat helpful with an above average level of satisfaction (Likert scale score: Exercise Tracker-3.00, iFitnessHero-3.73, iTreadmill-4.29).<sup>66</sup> Additionally, one study reported increased self-weighing frequencies for the patients (2.1 vs. 1 times/week) when compared to controls.<sup>67</sup> (See [Table 3.](#))

## Cost Efficacy

Only one study evaluated cost-effectiveness of Sense Wear Armband, a mobile armband device that can be used to provide weight loss intervention, concluding that the technology-based approaches were more cost effective and efficacious than traditional approaches in promoting weight loss via lifestyle changes in sedentary, overweight, and obese adults. The estimated costs per participant per kg lost for each group was \$129.15 for Group Weight-Loss education group, \$51.43 for Sense Wear Armband, and \$55.42 for group using both. This cost for Group Weight-Loss education group was approximately \$60 more than for the Sense Wear Armband alone.<sup>68</sup>

In a systematic review of technological components of weight management programs such as SMS, websites, and smartphone apps, efficacy data for only four apps (FitBit, iStepLog, My Meal Mate and Weight Watchers Mobile) out of the 22 analyzed were available and of these, FitBit One counted steps reliably, iStepLog increased likelihood to log steps, users were more adherent to My Meal Mate app than website or paper methods with overall weight loss of 4.6 kg over six months and Weight Watchers Mobile app + comprehensive program users were eight times more likely to reach their goal of 5-10 percent weight loss than those in the self-help group. All these apps were free to download.<sup>69</sup>

Another study reported extremely affordable SMS costs of average 10¢ per message sent and received.<sup>70</sup> In the course of the Text4Diet trial, participants were reimbursed for SMS costs of \$10/month. The trial also explored willingness of the participants to pay for the intervention and found that 89 percent were willing to pay \$4.99 for the intervention.<sup>71</sup>

Behavioral Weight Loss treatment (BWL) has traditionally involved long individual or group treatment sessions needing considerable time and resource investment from the patients resulting in restricted access outside research settings. The costs to implement behavioral weight-loss and lifestyle-change program for individuals with serious mental illnesses in community settings for a 12-month (30-session) intervention were reported to be \$16,427 or \$1095 per participant.<sup>72</sup> In a pilot study, Health-E-Call treatment was proposed as an alternative to the expensive BWL in an attempt to diminish costs and barriers to it.<sup>73</sup> Mobile Pounds Off Digitally (another BWL alternative) was reported as a remotely deliverable, low-cost (free to download podcast and apps) intervention with minimal participant contact requirement.<sup>74</sup>

Relatively low execution costs were reported for the Mobile Audio Manager and Response Tracker system because additional user creation and the supporting software were free, and the system could be used on any device.<sup>75</sup> In the Smart Coach for Lifestyle Management trial, the authors reasoned that face-to-face counseling supplemented with mobile technology was cost-effective because they noted weight loss levels comparable to that of the intensive counseling/smartphone group in the less-intensive counseling/smartphone group.<sup>76</sup>

## Discussion

Although a clear majority of the studies examining effects of obesity-related mobile apps on body weight found that the intervention group lost more weight than the control group did, this finding was not universal. Smartphone-enabled mobile apps appeared to be more effective than PDA-enabled or armband apps, although only a single study evaluated the effectiveness of each of the latter two types of apps. Considering the now nearly ubiquitous nature of smartphones in the United States (there are an estimated 222.9 smartphone users in the United States in 2016, a figure that is projected to rise to 264.3 million by 2021<sup>77</sup>), these positive findings regarding smartphone-enabled apps for obesity management are viewed as a positive sign for the future.

Effects of obesity-related apps on lifestyle behaviors were mixed, and the fact that such behaviors are surrogate measures of effectiveness must be kept in mind. The effects of mHealth apps on waste circumference and BMI were generally positive, as were the results measuring changes in dietary behavior as an output measure, but the variety of actual metrics used to measure changes in dietary behavior (e.g., consumption of consumption of fruits and vegetables, achievement of personal calorie goals, changes in daily caloric intake) make comparisons of the different studies problematic at best. Similar concerns existed with studies in which the outcome variable was changes in physical activity, especially because of the fact that increased physical activity, such as in the form of increased step count, was not necessarily associated with weight loss. This finding may be due to the general unreliability of most of these apps to count steps accurately.

Overall, the effects of obesity-related apps on patient satisfaction and adherence were positive and cost effective, and costs associated with the purchase and use of obesity-related apps were generally, but not always, relatively low, especially when compared with the costs associated with conventional behavioral weight loss therapy.

This literature review aimed to analyze results of all trials utilizing mobile technology directed at overweight and/or obese populations concerning their effectiveness and cost in comparison to conventional care. Consistent evidence suggested that mobile-based technological interventions have been efficacious in leading to changes in weight, BMI, waist circumference, and lifestyle behavior.

Lifestyle behaviors related to diet, physical activity, and sedentary behavior were mainly targeted. The interventions measuring changes in dietary intake and dietary behavior revealed an increase in fruit and vegetable intake and positive changes in eating behavior in all participants using the interventions under study. Overall, 80 percent of the studies that investigated caloric intake demonstrated a decrease in daily caloric intake. Furthermore, with these interventions, an increase in physical activity in the form of daily steps and exercise ranging from low to vigorous in intensity was observed in more than half of the studies.

Among the studies reviewed, we found minimal discussion or evaluation of the costs relative to the benefits of mHealth obesity interventions, with the exception of one study that found the estimated cost per participant per kilogram lost via technology-based approaches to be lower than the cost of traditional approaches. Few studies of the efficacy of mHealth interventions



mentioned associated costs or made inferences regarding the inexpensiveness of their interventions without citing the actual cost or running a cost-effectiveness analysis. Many of the apps included in these studies were offered for free, while some had minimal costs associated with the SMS component of the intervention. A large portion of the mobile apps analyzed were free or inexpensive, permitting easy access to the general population.

This literature review was limited by the design and quality of included studies, the number of databases accessed, and the search strategy utilized. It was difficult to identify the potential impact of mHealth on obesity-related measures because of the wide variation in study design. Also, with respect to quality, several studies were performed on small sample sizes and were of short duration, so inferences should be drawn with caution. Although some studies did not find clinically significant results during their short trial duration, the possibility of a longer-duration intervention producing possible significant results cannot be ruled out. Another main concern was that the publications on the cost-effectiveness of mHealth interventions for adult obesity were sparse, which restricts the generalizability of the findings. Furthermore, mobile device use necessitates that the study population possess a certain education level and socioeconomic status. Additionally, as the studies were evaluated to establish relevancy, publication and research bias cannot be ruled out.

In general, few studies addressed the cost-effectiveness of mHealth technologies. Most studies conducted so far have focused on assessing the quality of the outcomes, and these studies should continue, especially considering the quickly evolving nature of market offerings. However, in order to assess their usability and value for the money, further studies analyzing the cost-effectiveness of mobile health interventions are warranted. Additionally, whether a combination of conventional and mobile interventions for obesity would offer greater benefit for the cost, compared with a mobile-only approach, needs to be further studied.

A 2016 presentation at the European Obesity Summit indicated that nearly 29,000 weight management apps were currently available online from five mobile app stores (the Apple store, the Google store, the Amazon store, the Blackberry store, and the Windows store), but only 17 of them (less than 0.5 percent) had been developed using any kind of verifiable professional expertise.<sup>78</sup> Not only are better designed obesity-related apps needed, but many more well-designed studies are necessary to evaluate the market offerings in this fast-changing field.

## Conclusion

In this study, mHealth interventions were found to be more effective than conventional approaches in facilitating behavior modifications to promote weight loss and lifestyle changes to prevent and control adult obesity. Available data on concomitant expenses seem to indicate that these interventions can potentially be remotely delivered relatively inexpensively, improving access, promoting healthcare efficiency, and reducing the healthcare cost burden.

**Table 1: Mobile Interventions for Obesity: Study Characteristics**

Authors (Year)	Design	Theoretical Models	Sample Size	Participants	Trial length	Intervention /Control	Technological Components
<a href="#">Fukuoka et al.</a> (2010)	SGPP	NM	41	Sedentary women, mean age 48.4	3 weeks	Mobile apps	SM, feedback, goal setting
<a href="#">Archer et al.</a> (2012)	RCT	NM	197	GP, 18–65 years	9 months	a. GWL b. GWL + SWA c. SWA d. Standard Care	SM
<a href="#">Rabin et al.</a> (2011)	DD	NM	15	GP, mean age 38.9 years	1 week/app	Mobile apps	NM
<a href="#">Turner et al.</a> (2011)	RCT	SCT	96	GP, 18–60 years, BMI 25–45	6 months	a. Podcast b. Podcast + mobile	SM
<a href="#">Burke et al.</a> (2012)	RCT	SCT	210	GP, ≤59 years, BMI 27–43	24 months	a. PDA + feedback b. PDA	SM, CFC, SP



						c. Paper diary	
<a href="#">Allen et al. (2013)</a>	RCT	SCT	68	GP, 21–65 years, BMI 28–42	6 months	a. Intensive counseling b. Intensive counseling + smartphone c. Less-intensive counseling + smartphone d. Smartphone	SM, CFC, SP
<a href="#">Kim et al. (2013)</a>	RCT	NM	45	Older African American adults, mean age 69.3	6 weeks	a. SMS + pedometer + manual b. Pedometer + manual	SM
<a href="#">King et al. (2013)</a>	RCT	SCT; Social influence theory; Operant conditioning principles	68	GP, mean age 59.1	7 weeks	a. App + feedback b. App + group communication c. App + positive reinforcement	SM, problem solving, goal setting
<a href="#">Napolitano et al. (2013)</a>	RCT	SCT	52	Students, 18–29 years, BMI 25–50	8 weeks	a. Facebook + SMS + personalized feedback b. Facebook c. Wait-list control	SM, goal setting, CFC, ITP
<a href="#">Norman et al. (2013)</a>	RCT	SCT, control theory, operant conditioning	65	GP, 25–55 years, BMI $\geq 25.0$ –39.9	4 months	a. SMS b. Usual care	SM
<a href="#">Shapiro et al. (2012)</a>	RCT	SCT	128	GP, 21–65 years, BMI 25–39.9	12 months	a. SMS + MMS b. Monthly e-newsletters	SM, stimulus control tips, goal setting, problem solving
<a href="#">Spring et al. (2013)</a>	RCT	NM	69	GP, mean age 57.7	6 months	a. Mobile b. Standard care	SM
<a href="#">Steinberg et al. (2013)</a>	RCT	Self-regulation theory	91	GP, 18–60 years, BMI 25–40	6 months	a. SMS + usual care b. Usual care	SM, goal setting
<a href="#">Thomas et al. (2013)</a>	Pilot study	Self-regulation theory	20	GP, 18–70 years, BMI 25–50	12–24 weeks	Mobile app	SM, feedback
<a href="#">Bond et al. (2014)</a>	Experimental	NM	30	GP, 21–70 years, BMI $\geq 25$	4 weeks	App-prompted physical activity breaks + feedback + SWA arm band	SM, goal setting, feedback
<a href="#">Laing et al. (2014)</a>	RCT	NM	212	PCP	6 months	a. Mobile app b. Usual care	SM, goal setting, feedback
<a href="#">Cowdery et al. (2015)</a>	RCT	Self-determination theory	40	GP, 18–69	12 weeks	a. Mobile apps + PA tracker b. PA tracker	goal setting, monitoring

<a href="#">Gilliland et al. (2015)</a>	Quasi-experimental	Behavioral economic theory	208	GP, 18% BMI >30	8–10 weeks	SMS + up-front surveys + phone interviews + follow-up surveys + Google analytics data	SM, ITP
<a href="#">Svetkey et al. (2015)</a>	RCT	SCT and TTM	365	GP, 18–35 years, BMI $\geq 25$	24 months	a. Interactive cell phone app b. Personal coaching + selfmonitoring c. Control	SM, feedback
<a href="#">Sze et al. (2015)</a>	Pilot study	TAM	20	Parents (90% female), age 44.1 (SD 7.8) years; BMI 34.2 (SD, 6.8)	4 weeks	a. MAMRT + NI + EFT b. MAMRT + NI	SM, feedback, SP, ITP
<a href="#">Willey et al. (2016)</a>	Quasi-experimental	NM	10	Females, 30– 50 years, BMI 26.6–34	12 weeks	Mobile app	SM, feedback, SP, ITP

*Abbreviations:* BMI, body mass index; CFC, counselor feedback and communication; DD, Descriptive Design; EFT, Episodic Future Thinking; GP, general population; GWL, group weight-loss education; ITP, individual tailored program; MAMRT, Mobile Audio Manager and Response Tracker; ; MMS, Multimedia Messaging Service; NI, nutrition information; NM, not mentioned; PCP, primary care patient; RCT, randomized controlled trial; SCT, social cognitive theory; SGPP, single group pre- and post-intervention; SM, self-monitoring; SP, smartphone; SWA: Sense Wear Armband; TAM, technology acceptance model; TTM, transtheoretical model.

**Table 2: Distribution of the Participants in the Included Studies by Gender and Race**

Authors (Year)	Participant Characteristics			
	Gender	Percentage	Race/Ethnicity	Percentage
<a href="#">Fukuoka et al. (2010)</a>	F	100%	Minority	59%
<a href="#">Rabin et al. (2011)</a>	GP		NR	
<a href="#">Turner et al. (2011)</a>	F	48%	Non-Hispanic White	75%
			Non-Hispanic Black	19%
			Other	6%
<a href="#">Archer et al. (2012)</a>	F	81.7%	NR	
Burke et al. (2012)	F	84.8%	Non-Hispanic White	78.1%
<a href="#">Allen et al. (2013)</a>	F	78%	Non-Hispanic Black	49%
<a href="#">Kim et al. (2013)</a>	F	80.4%	Non-Hispanic Black	
<a href="#">King et al. (2013)</a>	F	73.5%	Non-Hispanic White	69%
			Hispanic/Latino	13%
			Asian	12%
<a href="#">Napolitano et al. (2013)</a>	F	86.5%	Non-Hispanic White	57.7%
			Non-Hispanic Black	30.8%
			Hispanic	5.8%

			Asian	1.9%
			Biracial/Multiracial	1.9%
			Other	1.9%
<a href="#">Norman et al.</a> (2013)	F	80%	Non-Hispanic White	75%
<a href="#">Shapiro et al.</a> (2012)	F	65%	Non-Hispanic White	64%
<a href="#">Spring et al.</a> (2013)	M	85.5%	Non-Hispanic White	75.4%
			Non-Hispanic Black	24.6%
<a href="#">Steinberg et al.</a> (2013)	F	75%	Non-Hispanic White	74%
<a href="#">Thomas et al.</a> (2013)	F	95%	Non-Hispanic White	85%
			Non-Hispanic Black	5%
			American Indian	5%
<a href="#">Bond et al.</a> (2014)	F	83.3%	Non-Hispanic White	66.7%
			Non-Hispanic Black	13.3%
			American Indian	3.3%
			Asian	3.3%
			Other	13.3%
<a href="#">Laing et al.</a> (2014)	F	73%	Hispanic or Latino	33%
			Non-Hispanic White	48%
			Non-Hispanic Black	19%
			Asian	8%
			Native American/Pacific Islander	2%
<a href="#">Cowdery et al.</a> (2015)	F	85%	Non-Hispanic White	85%
<a href="#">Gilliland et al.</a> (2015)	F	66%	NR	
<a href="#">Svetkey et al.</a> (2015)	F	69.6%	Non-Hispanic White	56.2%
			Non-Hispanic Black	36.2%
			Other	7.7%
<a href="#">Sze et al.</a> (2015)	F	90%	Non-Hispanic White	90%
			Other	10%
<a href="#">Willey et al.</a> (2016)	F	100%	Non-Hispanic White	
<i>Abbreviations:</i> F, female; M, male; GP, general population; NR, not reported.				

**Table 3: Effect of Mobile Health Interventions on Different Study Parameters**

Authors (Year)	Study Parameters					
	Body Weight (kg)	BMI (kg/m <sup>2</sup> )	Waist Circumference (cm)	Physical Activity Behavior	Dietary Behavior	Effects on Adherence/Satisfaction
<a href="#">Fukuoka et al.</a> (2010)	NE	NE	NE	Daily total steps increased by 800 or 15% over three weeks, SS	NE	Self-reported willpower improvements associated with step count increases

<a href="#">Archer et al. (2012)</a>	NE	NE	NE	NE	NE	Participants found all three apps easy to use, somewhat helpful with an above average level of satisfaction
<a href="#">Rabin et al. (2011)</a>	IG: 2.7%, CG: 2.7%, NS	IG: 0.91, CG: 0.26, SS	NE	Exercise self-monitoring rates (days/week) , SS: IG: 2.6, CG: 1.2) Intentional physical activity (kcal/day), SS: IG: 196.4, CG: 100.9	Decrease in daily calorie intake (kcal/day), NS: IG: 288.8, CG: 242.5 Self-reported diet (days/week): No differences	IG: App use to monitor diet 3.5 times more likely, more downloads per episode. CG: Majority used the website or paper.
<a href="#">Turner et al. (2011)</a>	GWL and SWA: 6.59, SWA: 3.55, GWL: 1.86, CG: none; SS	NE	NE	NE	NE	NE
<a href="#">Burke et al. (2012)</a>	PDA feedback: 2.32%, SS; PDA: 1.38%; CG: 1.94%; group differences NS	NE	PDA feedback: 6.4%, PDA: 5%, CG: 4%, SS	NE	NE	IG: Greater weight loss in participants who were adherent >60% of the time
<a href="#">Allen et al. (2013)</a>	IG: 2.3, CG: 0.63, NS	NE	NE	NE	IG: Positive effects on fruit and vegetable intake and eating behavior inventory scores	Response to messages: All (first week), two of three texts (by week 16)
<a href="#">Kim et al. (2013)</a>	IG: 1.65 (1.8%), CG: 1.03 (0.8%), NS	NE	NE	IG: Steps increased to 3,000 steps/day, SS	NE	Text-messaging adherence moderately strong (60% – 69%). Greater weight loss in adherent participants.
<a href="#">King et al. (2013)</a>	IC + SP: 5.4, LIC + SP: 3.3, SS	IC: 0.8, IC + SP: 1.8, LIC + SP: 1.1, SP: 0.7, SS	IC: 3.0 (M), 3.19 (F); IC + SP: 7.01 (M), 5.68 (F); LIC + SP: 6.5 (M), 3.64 (F); SP: 3.38 (M), 0.88 (F), SS	Self-reported physical activity of moderate or greater intensity (hours/week) decreased in all groups (IC: -1.4, IC + SP: -2.0, LIC + SP: -3.6) except for slight increase in SP group (0.19)	Decrease in daily calorie intake (kcal/day), SS: IC: 415.6, IC + SP: 468.2, LIC + SP: 218.5, SP: 249.2 Increase in fruit and vegetable servings/day, NS: IC: 0.81, IC + SP: 0.51, LIC + SP: 2.1, SP: -0.05	Adherence highest in IC + SP group: Counseling attendance: 72% App diet logging: 53% App physical activity logging: 32% Similar percentages for LIC + SP group

<a href="#">Napolitano et al. (2013)</a>	NE	NE	NE	Increase in daily steps: IG: 679; CG: 398 Changes in MET: IG: 11.96; CG: 4.55	NE	NE
<a href="#">Norman et al. (2013)</a>	NE	NE	NE	Increases in brisk walking for all groups (IG 1: 71.1, IG 2: 122.9, IG 3: 105.7) and MVPA for all groups (IG 1: 179.2, IG 2: 257.1, IG 3: 134.3), SS	NE	Participants continued to use the applications for 211.0 days (IG 1), 199.3 days (IG 2), and 162.0 days (IG 3) after the study period
<a href="#">Shapiro et al. (2012)</a>	Facebook Plus: 2.4, Facebook: 0.63 CG: 0.24, SS	NE	NE	Differences within or among the groups: NS	Differences within or among the groups: NS	97% found the program helpful and 100% would recommend the program
<a href="#">Spring et al. (2013)</a>	IG: 2.9, CG: 0.02 (gain), SS	NE	NE	NE	NE	Adherent participants in the IG lost more weight than adherent or nonadherent participants in the control group
<a href="#">Steinberg et al. (2013)</a>	IG: 7.7 (6.55%), CG: 1.5 (0.35%), SS	IG: 0.47, CG: 0.42, NS	NE	NE	IG participants self-weighed more days/week and consumed fewer calories/day than CG participants Daily caloric intake (kcal/day), SS: IG:1, 509, CG:1,856	IG: daily self-weighing was perceived positively
<a href="#">Thomas et al. (2013)</a>	IG: 10.9 kg, no CG, SS	NE	NE	NE	NE	Adherence to self-monitoring: 85% All extended program participants gave the maximum rating for satisfaction and would recommend the program
<a href="#">Bond et al. (2014)</a>	NE	NE	NE	SED decrease, SS, 3 -min: 5.9%, 6- min: 5.6%, 12-min: 3.3% LPA increase, SS, 3 -min: 3.9%, 6-min: 3.9%, 12-min: 1.9% MVPA increase,	NE	App use increased motivation to take PA breaks and reduced SED

				SS, 3 -min: 2.0%, 6- min: 1.7%, 12- min:1.3%		
<a href="#">Laing et al.</a> (2014)	IG: 0.30, CG: 0.27, NS	NE	NE	Self-reported exercise behaviors: difference NS	IG used “personal calorie goal” more often than CG did (group difference: 2.0 days/week)	Frequency of logins (mean total logins: 61) decreased rapidly after enrolment. No association between baseline characteristics and extent of app use or weight change.
<a href="#">Cowdery et al.</a> (2015)	NE	IG:0.09 21, CG: 0.1879 (gain), NS	NE	Decrease in total MET (min/wk), NS: IG: 356.8 CG: 722	NE	NE
<a href="#">Gilliland et al.</a> (2015)	NE	NE	NE	NE	80% benefited Greater participation associated with increased healthy eating. No influence on consumption behavior.	NE
<a href="#">Svetkey et al.</a> (2015)	CP: 0.99, PC: 2.45, CG:1.44, NS	NE	NE	NE	NE	Self-weighing frequency(times/wk): CP: 4.0 (0–6 months), 2.1 (13–24 months) PC: 2.2 (0–6 months), 1.0 (13–24 months) All monthly calls completed: >90% (0– 6 months), >87% (13–24 months)
<a href="#">Sze et al.</a> (2015)	% overweight: IG: 4.6, CG: 1.1, SS	IG: 1.0, CG: 0.2, SS	NE	NE	Decrease in calorie intake (kcal/day), SS: IG: 791.2, CG: 482.4	IG: 87%, CG: 92.9%
<a href="#">Willey et al.</a> (2016)	IG: 6.13 (7.3%), no CG, SS	NE	IG: 7.2 (6.6%), no CG, SS	NE	NE	More app use associated with greater improvement in each health outcome but results were NS

*Abbreviations:* BW, body weight; BMI, body mass index; CG, control group; CP, cell phone; F, female; GWL, group weight-loss education group; IC, intensive counseling; IG, intervention group; LIC, less-intense counseling; LPA, low-intensity physical activity; M, male; MVPA: moderate to vigorous intensity physical activity; NE: not evaluated; NS, nonsignificant; PA, physical activity; PC, personal coaching; SED: excessive sedentary time; SP, smartphone; SS, statistically significant; SWA: Sense Wear Armband.

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