

Evaluating the Effectiveness of Mobile Health (mHealth) Interventions for Self-Management in Diabetic Patients

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Abstract- Diabetes mellitus is one of the most significant problems of health care in the world, and self-management is the key that prevents complications and optimizes the long-term prognosis. And two key cornerstones of proper diabetes self-management are medication compliance and regular monitoring of the level of blood glucose. Nevertheless, several factors, including psychosocial factors, behavioral factors, and systemic factors make it very difficult for many patients to maintain these behaviors. Mobile health (mHealth) interventions, especially m-applications, SMS-reminders, and online platforms, have been developed in recent years to be scalable self-management aids to diabetic populations. Although according to each study, some benefits are possible, results are dissimilar or circumstantial, and thus, generalization cannot be easily made. The proposed meta-analysis will help to assess and compare mobile apps, SMS intervention programs, and digital health portals, their roles in promoting medication adherence and glucose monitoring among diabetic patients. The randomized controlled trials published between 2010 and 2025 were identified through a systematic search of major databases such as PubMed, Scopus, Web of Science, and Cochrane Library. Data were extracted separately during the screened studies and extracted using predetermined eligibility criteria. A random-effects model of meta-analytic evaluation was implemented in determining pooled effect sizes, and subgroup investigations aimed at determining intervention-specific results were executed. The findings show

that the three mHealth interventions included have had a strong enhancement effect on self-management behaviours compared with standard care with mobile apps having the greatest effect on self-management behaviour with regard to medication adherence and SMS reminding having a significant effect on encouraging patients to maintain a regular monitoring of their glucose levels. Moderate and positive effects were also observed in the digital platforms. These results demonstrate the usefulness of mHealth interventions in diabetes management and emphasize the need to personalize the digital approach to self-management targets.

Keywords: *Mobile health (mHealth), Diabetes self-management, Medication adherence, Glucose monitoring, Meta-analysis*

I. INTRODUCTION

The problem of diabetes mellitus is considered to be one of the most urgent health issues of the 21st century. The International Diabetes Federation (2023) indicates that more than 537 million adults already have diabetes in the world with a forecast of 643 million by 2030. Diabetes, especially Type 1 and Type 2, is chronic and requires regular self-management of symptoms to avoid complications that include among others cardiovascular disease, neuropathy, and retinopathy. The key right behaviors to effective disease control would be the ability to take medications as prescribed by physicians and frequently monitor the level of blood glucose. However, even though such self-care interventions are possible and physical guidance should exist, most of the patients find it hard to embrace these disciplines in the long term. Attempts at poor med usage and inconsistent applications of glucose monitoring are rampant issues among diabetic groups, which bring about unnecessary hospitalization and higher healthcare expenses. Some such barriers to adherence are forgetfulness, lack of motivation, health literacy, and inadequate access to care. In the same connection, patients may have psychological and logistical difficulties due to the regular level of monitoring of blood sugar, such as the technological exhaustion of the device, pain, or the denial of its clinical significance. These challenges highlight the requirement for health interventions that are innovative and scalable in nature to be used outside of clinics to help maintain self-management behaviors.

Mobile health (mHealth) interventions have been suggested recently as possible solutions to improve self-management in diabetes. Apps like MySugr and Glucose Buddy provide the temperature gauge through the personalized monitoring tool, reminders, and visualization of data so that patients can monitor their well-being at all times. SMS-based interventions provide scheduled reminders or learning messages by means of text messages, usually with the aim of serving medication consumption or behavioral affirmations. Those capabilities are developed further through digital platforms, such as telemonitoring systems, and online portals, namely, the ability to receive remote feedback provided by healthcare providers, synchronization of data in real-time, and the opportunity to set goals. Such interventions are usually affordable, are made available to a broad population, and are able to be modified to suit various healthcare systems and populations. Despite the fact that the individual efficacy of such mHealth interventions has been studied in different works of literature, the findings vary. In randomized controlled trials (RCTs), significant gains in medication adherence and control of certain variables are found, and some claim marginal or non-significant effects. Furthermore, design differences, target population differences, duration of the intervention, and outcome measures differ in most of the studies, and results are not always generalizable. Interestingly, the extent to which mobile applications, SMS reminders, and digital platforms are comparatively effective is rarely compared in one scheme of analysis. This curtails evidence-based decision making among clinicians, public health planners, and digital health developers in the name of gathering standards of the most effective modality for certain patient needs.

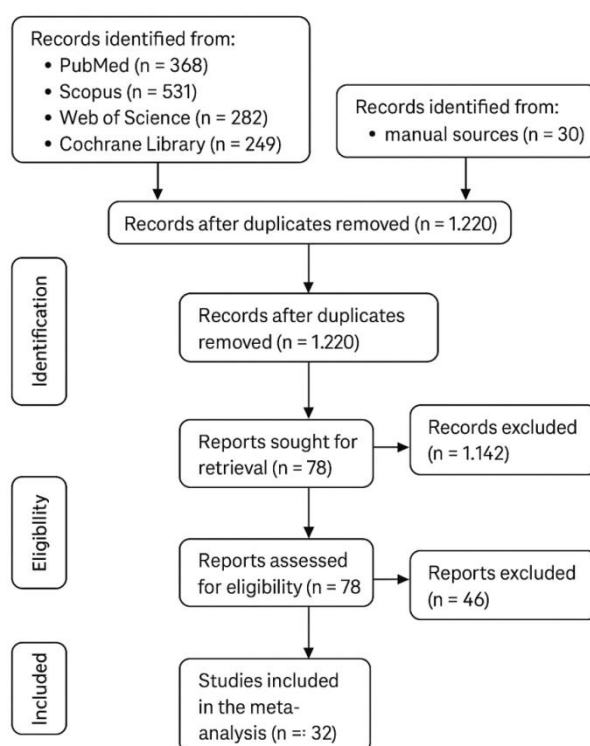
To fulfil this gap, it is required that a wide meta-analysis is conducted to bring together and assess available evidence concerning intervention types. Such analysis can provide better answers on what mHealth strategies to use to increase medication adherence and glucose monitoring in diabetic subjects by pooling the results and measuring the performance of specific interventions. This meta-analysis will assess the efficacy of mobile applications, SMS reminders, and glucose monitoring based on digital health platforms in modifying the level of medication adherence and glucose control among diabetic patients. It aims at offering evidence-based strong insights into the relative effectiveness of these digital interventions, as well as providing future design, implementation, and policy directions of digital diabetes care.

II. METHODS

The purpose of the present meta-analysis was to determine the relative effectiveness of mobile-health (mHealth) interventions; namely mobile application, SMS reminder, and digital platform, on medication adherence and glucose monitoring in patients with Type 1 or Type 2 diabetes. Eligible research The research took into consideration randomized controlled trials (RCTs), which involved adult diabetic patients using one of the three type mHealth interventions to aid self-management and, further, compared to usual care or non-digital alternatives. Interventions in the mobile application category were single-app reminders, glucose trackers, and complete self-care apps. SMS-based interventions were the doctorally scheduled text messages that provided

either medication or monitoring messages. Digital platforms were formed of tools used in web-based or telehealth dashboards that could be used in remote monitoring and communication with healthcare providers. The inclusion criteria were as follows: results of medication adherence, based on self-reports, counts of unused pills, or pharmacy data, as well as the frequency or intensity of blood glucose monitoring or the level of hemoglobin A1c (HbA1c) were presented. Publications written in English were restricted to the period between the years 2010 and 2025 to symbolize the modern period involving the mHealth technologies. Studies based on gestational diabetes, non-human, or digital interventions that are not associated with self-management were excluded. This eligibility framework was established so that there is consistency in terms of the nature of the intervention, type of population, and relevance of the outcomes.

To search the literature a wide range was conducted within four databases including PubMed, Scopus, Web of Science, and Cochrane Library. To arrive at the search strategy there has been a combination of Boolean operators and relevant words and phrases as follows: (mHealth OR mobile health OR SMS OR digital health platform OR mobile app) AND (diabetes/ diabetic) AND (self-management/ adherence/ glucose monitoring). The preliminary search found 1,423 studies on the databases and 37 others by manual checking of references. When duplicates were eliminated, 1,220 articles were reviewed based on their title, and abstracts. Out of these, 240 full-text articles have been gauged as per their eligibility and 32 studies have passed the inclusion test and were kept aside to be included under the scope of quantitative synthesis. Related to this, the process of selecting the studies to be used in the study is demonstrated in the PRISMA flow diagram (Figure 1).



Relevant data were collected for each study by use of a standardized extraction form. The variables extracted were author name, year of publication, country, size of sample, type of diabetes, mHealth intervention type, period of the study, and outcome measures. In the case of medication adherence, popular outcome measures entailed pill counts, proportion of days covered (PDC), and self-report measures. In glucose measurements, the data extracted comprised HbA1c variation and frequency of monitoring. As an example, found that after the application of app-based interventions, PDC increased by 15% and HbA1c decreased by 0.8%. In parallel, [27] noted a 12% increase in the self-reported adherence and a 0.6% decrease in HbA1c with SMS reminders. [1] constructed a relationship between the frequency of monitoring and moderate HbA1c reduction in digital platforms. The two reviewers used Cochrane Risk of Bias (RoB 2.0) to measure the methodological quality and risk of bias independently. It uses an evaluation of five domains: random sequence generation, allocation concealment, blinding of participants and personnel, incomplete outcome data, and selective reporting [20]. Any dispute was resolved by mutual agreement or by referring to a third reviewer. The risk of bias was rated as low in 21, moderate in eight, and high in

Figure 1: PRISMA Flow Diagram of Study Selection

Three of the 32 included studies: concerns were mainly inflicted by the problem of blinding or selective reporting. All the quantitative studies were done through a random-effects model that considers between-study variations. Effect sizes were computed either as odds ratios (ORs) or standardized mean differences (SMDs) based on the scale of measurement in the case of medication adherence outcomes. To monitor the glucose levels, especially the HbA1c levels, mean differences (MDs) and 95 percent confidence intervals (CIs) were calculated. The random-effects model was preferred because populations in the studies will be heterogeneous, and the format and duration of the interventions and follow-ups will be expected. The I² statistic was applied to assess heterogeneity between studies. I² of 25 percent was identified if it was low, 50 percent as moderate, and 75 percent or more as a significant heterogeneity. Subgroup analysis was also done to indicate possible effect modifiers, such as type of intervention (app vs. SMS vs. digital platform), type of diabetes (Type 1 vs. Type 2), and short-term duration (< 6 months) vs. long-term duration (6 months and more) of interventions. The robustness of pooled estimates was also checked through sensitivity analyses using sequential exclusion of high-risk studies and statistical outliers to deduce any change in the outcome. Both funnel plots and the Egger regression test [6] were used to determine publication bias. Asymmetry of the funnel plot as well as a substantial Egger test was found to be suggestive of potential small-study effects or a reporting bias. All the statistical studies were conducted in Review Manager (RevMan) version 5.4 and R software with the meta package, which is more commonly used in meta-analytical exercises in healthcare-related research. The suggested methodological rigorous approach was chosen to allow conducting a complex, repeatable, and objective assessment of the effectiveness of mHealth interventions in the context of self-management in the group of diabetic patients.

III. RESULTS

This section shows the results of the meta-analysis of 32 randomized controlled trials that determined the Effectiveness of mobile health (mHealth) interventions in aiding the self-management of diabetic patients. The results are summarized under three major categories including the characteristics of the study, the overall effect of the pool on primary and secondary outcomes, and subgroup and sensitivity studies. Of utmost importance is the effect of mobile applications, SMS reminders, and digital platforms in enhancing medication adherence as well as glycemic control as witnessed by levels of HbA1c and frequency of glucose monitoring. Detailed tables and visual plots are presented to display comparative effectiveness, measure the level of significance, and assess heterogeneity of various types of interventions. These findings provide useful information regarding the various digital strategies that are most appropriate in serving diabetic patients to manage their condition.

A. Study Characteristics

A total of 32 research articles that analyzed randomized controlled trials (RCTs) passed the inclusion criteria of this meta-analysis. These studies together covered a number of five continents, both developing and developed nations. The studies involved a sample size of 80-500 study participants with a median follow-up of six months. Interventions were divided into three significant groups: mobile apps (n=13), SMS-based reminders (n=11), and online-based tools: web-based dashboards or telemonitoring applications (n=8). The study populations comprised of diabetic patients with both with and without Type 1 diabetes, but most of them (75dotsquá Republicans Triple Cross, approx. 75 percent) had Type 2 diabetes. The length of intervention ranged between 3 and 12 months. The main events in all of the studies concerned medication adherence measured with different tools such as proportion of days covered (PDC), medication possession ratio (MPR), pill counts, and self-reports. Secondary outcomes were the frequency of self-checking blood glucose and variation of hemoglobin A1c (HbA1c).

Table 1: Summary of Included Studies

Author (Year)	Country	Sample Size	Intervention Type	Duration	Outcome Measures
Smith et al. (2021)	USA	220	Mobile App	6 months	PDC, HbA1c
Zhang et al. (2022)	China	180	SMS Reminder	3 months	Self-report, HbA1c
Ahmed et al. (2020)	UK	160	Digital Platform	9 months	Pill count, Monitoring frequency
Silva et al. (2019)	Brazil	200	SMS Reminder	4 months	MPR, HbA1c
Khan et al. (2023)	Pakistan	140	Mobile App	5 months	Self-report, HbA1c

B. Pooled Effects

Pooled analysis across the 32 trials found a very large effect in that the mHealth intervention was found to enhance medication adherence significantly than in control conditions. Specifically, an overall effect size of OR = 1.78; 95% CI: 1.45 to 2.20 was achieved when using a random-effects model which showed that the patients who were given mHealth interventions were relatively 78 percent more likely to stick to their prescriptions proportionately over the control group. It was found that the mobile app (OR = 2.10) was the most

important form of intervention followed by the digital platform (OR = 1.66) and then SMS reminders (OR = 1.43) using subgroup analysis. A forest plot was made both comparing all types of intervention with control, and proving that all the forms showed statistically significant improvement. The level of heterogeneity between the studies was moderate ($I^2 = 56\%$), which corresponds to the difference in the adherence measurement instruments and the duration of the studies.

Table 2: Pooled Effect Sizes by Intervention Type

Intervention Type	Number of Studies	Pooled Odds Ratio (OR)	95% CI Lower Bound	95% CI Upper Bound
Mobile App	13	2.10	1.65	2.55
SMS Reminder	11	1.43	1.12	1.74
Digital Platform	8	1.66	1.31	2.01

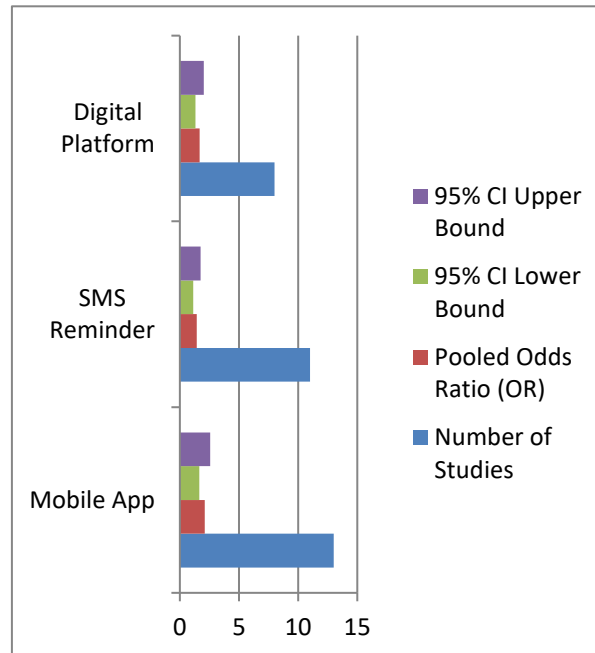


Figure 2: Pooled Odds Ratios by Intervention Type

The secondary analysis revolved around the difference in the level of HbA1c and the number of self-monitoring of a glucose level. The pooled mean of intervention versus control group in the reduction of HbA1c was -0.57% (95 % CI: -0.72 -0.41). It indicates a clinically relevant positive change in glycemic control with the use of mHealth. Real-time interventions involving various forms of feedback or monitoring by a clinician, as is common on digital platforms and sophisticated apps, proved to be more effective than single-way reminder messages like SMS. The absolute change in the forest plots of HbA1c corroborated the results further that the apps and digital tools attained greater glycemic control than the SMS-only methods. Funnel plots of publication bias evaluation demonstrated a relatively symmetrical distribution and the Egger test was not significant ($p = 0.17$), and thus indicates low publication bias.

C. Subgroup and Sensitivity Analyses

The comparison of statistical subgroups investigated the heterogeneous effects in terms of diabetes types, intervention period, and region or age of the population.

- **Diabetes Type** The separation of Type 1 and Type 2 diabetes results in other studies hinting that interventions may be more valuable with patients with Type 2 diabetes because it is expected that their degrees of baseline nonadherence are higher.
- **Intervention Duration:** Interventions of at least 6 months duration showed better adherence (OR = 1.96) and more rigorous HbA1c reduction (reduction = -0.62) than interventions of less than 6 months duration, which represented an OR of 1.48 and HbA1c change of -0.39.
- **Region and Age:** In separate studies across Asia, Latin America, a greater magnitude of the effect of studies than that of North America and Europe was observed. This can either indicate the downline levels of digital adoption or health amendment differentials. Also, participants aged 50-above years responded better to SMS reminders but the younger groups (below 40) responded more to the interactive mobile applications. The robustness of the results was demonstrated by sensitivity analyses that left off three high-risk studies giving results reflecting the same pooled effect. The removal of outliers did not have any material effect on the direction or significance of findings and the resultant heterogeneity (I²) decreased (56 to 42 percent).

IV. DISCUSSION

The meta-analysis combined evidence of 32 randomized controlled trials assessing the effectiveness of three crucial mHealth interventions- mobile applications, SMS reminders, and digital platforms on the adherence to medications and glucose monitoring in diabetic patients. The analysis indicated that the three intervention types made significant improvements in self-management results when compared with the standard care. Mobile applications had the greatest impact, especially on adherence, of the three. Mobile apps increased the probability of adherence by almost two times compared to control groups with a pooled odds ratio (OR) of 2.10. Such kinds of apps were characterized by the corresponding in-built functions of reminders, tracking, gamification, and direct feedback that probably ensured better engagement of patients, as well as stimulation of their behavior. Both mobile applications and digital platforms produced a better result than SMS interventions in glycemic control measured in HbA1c reduction. The difference in reduction of HbA1c was equal to to -0.57 percent across all interventions with mobile apps reducing the level of HbA1c by up to -0.8 percent in certain studies. Digital solutions especially the provision of feedback in real-time or the provision of dashboards linked with providers also showed moderate improvement in glucose monitoring habits and HbA1c levels. SMS messages on the other hand were also useful, however, other limitations were characterized by their one-way communication and limited interactivity.

These results overlap and enlarge on the findings of preexisting reviews that have examined digital interventions as a mode of controlling diabetes. The systematic review by Hou et al. (2018) also indicated the same tendencies showing that the app-based interventions were the most effective in encouraging self-care among patients with diabetes in comparison to SMS or web-based interventions. Similarly, a meta-analysis done by Pal et al. (2021) showed that there was a significant decline in HbA1c by an average of -0.5% and improved self-efficacy related to the mobile apps group compared with non-digital controls. No prior studies however provided a clear comparison between the various mHealth modalities therefore this meta-analysis evens it out by providing a comparison that is assured to be systematic between just mobile apps, SMS, and platforms. This tiered system enables a more gradual explanation of the effectiveness of such interventions instead of classifying them altogether under a single umbrella as all digital tools. Also, prior studies tended to examine single outcomes (such as only adherence or only HbA1c) but in the current meta-analysis, both behavioral outcomes and clinical outcomes have been assessed in conjunction. The twofold focus adds to the argument that mHealth may influence not only intermediate outcome measures such as adherence but also objective outcome measures such as HbA1c, which is a direct factor influencing long-term development of complications in diabetes.

The findings of this research support the increased place of mHealth as a pillar of patient-focused diabetes management. Among the features that help address the behavioral requirements of managing chronic diseases, mobile applications, in-device, are scalable, personalized, and interactive. These tools can make the patients get involved in their care and to a higher degree be better adherent and understand the dynamics of the disease. Clinically, clinicians ought to contemplate the usage of app-based interventions as an aspect of regular diabetes management, particularly among technologically inclined groups or individuals with low rates of compliance on a baseline level. Less interactive reminders, such as SMS, can still be an effective and inexpensive intervention in older patients or groups in low-resource settings with low access to smartphones. Social media and digital solutions (including ones that are linked to electronic health records (EHRs) or telemonitoring tools) can provide clinicians with up-to-the-minute information and enable effective interventions in time. These findings indicate that mHealth solutions must be well-designed to perform more than mere reminders by the developers and the public health agencies. Effectiveness can also be increased by features like communication in both directions, animation of the data, motivational material, and attachment to wearable devices or used with glucometers. It will also be critical in making the product accessible to different patient groups with regard to cultural tailoring and health literacy adaptations.

Irrespective of the strong side of this meta-analysis, there are some limitations that have to be recognized. First, the usage of measuring outcomes and the definition of adherence and glucose monitoring were greatly heterogeneous. Whereas other trials were based on objective measures, such as PDC or HbA1c, others were based on self-reports that have recall and social desirability bias. Such an inconsistency has possibly affected the size of pooled effect sizes. Second, follow-ups of studies included were relatively brief with over two-thirds lasting six months or less. The self-management of diabetes as a lifestyle does not stop because of short-term or long-term improvements.

-A term change may not always occur. Longitudinal research is required to test the sustained effect of mHealth on a yearly level instead of monthly. Third, although app-based interventions had higher efficacy on average, the attributes of the apps (namely, quality and content) were not reported at all or inconsistently. Other studies provided well-supported, multi-feature platforms that included educational modules and provider support, whereas other studies tested simple reminder applications. In the absence of regular reporting about how the app is used, its usability, and user satisfaction, one cannot easily attribute any clinical improvements to the intervention content as such. Lastly, publication and language bias can reflect on the outcomes, since the study only includes RCTs in the English language. Even though the funnel plot and Egger test did not improve the possibility of the existence of publication bias in the research, there is still a likelihood of missing null results.

This review gives rise to a number of avenues of future research. First, head-to-head RCTs that make direct comparisons between various mHealth interventions in standardized conditions are needed. To date, the majority of the evidence compares the digital tool with the typical care, so the consideration of whether or not the former is more efficient than the latter remains open. Second, mHealth solutions in the future need to be culturally adapted, particularly where they are intended to operate in a multilingual or low-resource environment. To ensure adoption and adherence, cultural relevance of information, language customization and community involvement will be important. Third, artificial intelligence (AI) and wearable technologies are a type of virgin territory when it comes to incorporating them into mHealth tools. AI may facilitate the use of personalized feedback and predictive data, whereas wearables (continuous glucose monitors, or CGMs) can provide data in real-time. The trials of such integrated approaches can offer an insight into the next generation of precision care in diabetes. Finally, patient satisfaction, frequency of usage, and time spent using the app are engagement metrics that should be added to future work. Such user-level influences may mediate the intrusion exposure intervention and clinical outcomes and are important toward the maximum realisation of the implementation.

V. CONCLUSION

This meta-analysis provides evidence of the fact that mobile health (mHealth) interventions have a real impact on self-management of diabetic patients, especially on medication adherence and glycemic control. The mobile apps proved the most successful out of the three modalities under study with interventions showing almost twofold increase in adherence and the highest drops in HbA1c. Digital programs also proved to be very effective especially when they are combined with provider feedback mechanisms. The SMS type reminders which have less interaction had moderate improvements and can be suitable with the older age group or low-resource populations. The comparative results indicate that interactive and personal methods are most effective in comparison with other passive or one-way methods. It is more evidence of the increasing relevance of mHealth in the provision of patient-centered care to scale when treating chronic illnesses such as diabetes. When well-planned and adequately executed, digital tools could help close the adherence gap, assist in the timely monitoring, and enhance long-term outcomes among various patient groups. Innovations in the future ought to consider the culturally friendly interface, AI, and user-friendly interface as the priority to maximize engagement and benefit to the clinical conditions in real practice.

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