



Leveraging Mobile Health Technologies for Improved Hypertension Control and Patient Adherence

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Abstract

Background: Hypertension is a major global health challenge and a leading risk factor for cardiovascular morbidity and mortality. Despite effective pharmacological and lifestyle interventions, suboptimal adherence to treatment and limited patient engagement remain significant barriers, especially in low- and middle-income countries. Mobile health (mHealth) technologies have emerged as promising tools to address these gaps by supporting adherence, enabling self-monitoring, and fostering behavior change.

Objective: This study aimed to explore the role of mHealth technologies in hypertension management, focusing on their impact on treatment adherence, patient engagement, and self-care behaviors.

Methods: A convergent parallel mixed-methods design was employed, combining a cross-sectional online survey (n = 150) and semi-structured interviews (n = 30). Participants were adults aged 30–70 years with primary hypertension who had used at least one mHealth application for a minimum of three months. Quantitative measures were guided by the Technology Acceptance Model (TAM) and Health Belief Model (HBM), while qualitative interviews captured lived experiences and contextual factors. Statistical analyses included t-tests, ANOVA, and logistic regression; qualitative data were analyzed using thematic analysis.

Results: mHealth usage was widespread, with 91% of participants using apps for blood pressure tracking, medication reminders, or lifestyle monitoring. High-frequency app users demonstrated significantly better adherence scores ($p < .01$) and more consistent blood pressure self-monitoring ($p < .001$). Qualitative findings highlighted empowerment through self-monitoring, integration into daily routines, and motivation via feedback and gamification. However, barriers such as technical difficulties, lack of personalization, privacy concerns, and digital inequities—particularly among older adults—limited sustained use.

Conclusion: mHealth technologies offer significant potential in improving adherence and self-management in hypertension care. Their effectiveness depends on user-centered design, cultural relevance, personalization, and equitable digital access. Future strategies should integrate behavioral frameworks, enhance digital literacy, and ensure inclusivity to maximize clinical impact.

Keywords: Hypertension, mHealth, Patient adherence, Self-management

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Introduction

Hypertension, or sustained high blood pressure, is one of the most pressing global health concerns and a leading risk factor for cardiovascular diseases (CVD), including stroke, myocardial infarction, heart failure, and renal failure. According to the World Health Organization (WHO), approximately 1.28 billion adults worldwide are living with hypertension, yet only 42% are diagnosed and treated, and fewer than one in five achieve adequate blood pressure control. As a long-term condition, hypertension contributes significantly to premature mortality, morbidity, and healthcare costs, representing not only a burden to individuals but also to health systems and national economies. Despite the availability of effective pharmacological therapies and lifestyle-based interventions, optimal management of hypertension remains challenging, particularly in low- and middle-income countries (LMICs), where limited healthcare access and continuity exacerbate the problem (1).

A major obstacle to hypertension control is poor patient adherence to prescribed medications and lifestyle changes. Research consistently demonstrates that adherence is influenced by a range of factors, including forgetfulness, medication side effects, absence of symptoms, insufficient disease knowledge, and low motivation (2). In addition, healthcare system-level barriers such as brief consultations, ineffective communication, and inadequate follow-up further compromise adherence (3; 4). Lifestyle modifications—including reduced salt intake, regular exercise, weight management, smoking cessation, and stress reduction—are well established as essential elements of hypertension management. However, long-term adherence to these behavioral changes remains limited, largely due to difficulties in maintaining motivation and engagement (5; 6).

Given these barriers, there is a clear need for scalable, cost-effective, and sustainable solutions that provide continuous patient engagement, support, and monitoring beyond the clinical setting. In recent years, mobile health (mHealth) technologies have emerged as promising tools to address these challenges. Broadly defined, mHealth refers to the delivery of healthcare information and services through mobile phones, wireless devices, and digital platforms (7). With the exponential growth in smartphone adoption across both developed and underserved regions, mHealth interventions represent an unprecedented opportunity to overcome healthcare disparities and extend care into patients' daily lives.

Evidence suggests that mHealth interventions can enhance patients' knowledge of their condition, support adherence to treatment, and promote sustained lifestyle changes (8; 9). Specifically, mobile applications designed for hypertension management can provide real-time blood pressure monitoring, automated medication reminders, interactive dashboards, educational resources, and motivational prompts. By tailoring interventions to user needs, mHealth solutions can foster patient autonomy, accountability, and engagement. Furthermore, integration with wearable devices and connected home monitors enables continuous data collection, offering patients and clinicians a more complete picture of blood pressure patterns and treatment responses (10). These features are particularly valuable for individuals with fluctuating blood pressure, white-coat hypertension, or comorbidities requiring close monitoring.

Despite their potential, the effectiveness of mHealth interventions is heavily dependent on usability, perceived value, and integration into daily life. Studies indicate that poorly designed applications—those with complex interfaces, lack of cultural relevance, or limited personalization—often lead to low engagement and high dropout rates (11; 12). To address this, usability testing tools such as the Mobile App Usability Questionnaire (MAUQ) have become essential for developing patient-centered digital health tools (13). Co-design approaches, which involve patients in the development process, have also been shown to improve relevance, acceptability, and sustained use of mHealth applications (14). Beyond usability, digital health equity is a critical concern.



Socioeconomic disparities, limited digital literacy, and poor internet access hinder the adoption of mHealth technologies, particularly among older adults, rural populations, and marginalized groups (15; 16). The COVID-19 pandemic further highlighted the digital divide, as many vulnerable populations were unable to access telehealth or mobile health services due to technological barriers and linguistic or cultural mismatches (17). Addressing these inequities requires comprehensive strategies, including infrastructure investment, digital literacy initiatives, and inclusive policy frameworks, to ensure that mHealth interventions do not widen existing health disparities.

The effectiveness of mHealth solutions can also be explained through behavioral science frameworks. The Health Belief Model (HBM) posits that individuals are more likely to adopt health-promoting behaviors when they perceive themselves at risk and believe in the effectiveness of recommended actions. Mobile applications can support this by providing real-time feedback and alerts, thereby reinforcing perceptions of risk and the importance of adherence. Similarly, the Theory of Planned Behavior (TPB) emphasizes the role of attitudes, subjective norms, and perceived behavioral control in shaping behaviors. mHealth interventions that incorporate social support functions, clinician messaging, and progress tracking can strengthen self-efficacy and motivation, ultimately leading to better adherence. Researchers argue that embedding such theoretical frameworks in mHealth design enhances their effectiveness and sustainability (8).

The literature provides encouraging evidence of the potential of mHealth in hypertension management. Systematic reviews and meta-analyses demonstrate that mobile interventions can significantly reduce both systolic and diastolic blood pressure and improve medication adherence, dietary habits, and physical activity levels (6). Importantly, these benefits extend beyond passive data collection; they actively foster behavior change, self-monitoring, and long-term engagement. However, challenges remain. Usability issues, cultural mismatches, and inequities in digital access continue to limit the reach and effectiveness of mHealth solutions.

In summary, hypertension remains a critical public health challenge, with poor adherence and limited patient engagement undermining effective management. mHealth technologies hold significant promise as scalable, cost-effective tools to address these barriers by enhancing adherence, enabling self-monitoring, and supporting behavior change. Nevertheless, their success depends on thoughtful design, inclusivity, and integration with behavioral theory. Future research and implementation strategies must therefore focus not only on demonstrating clinical effectiveness but also on ensuring accessibility, equity, and sustainability in diverse patient populations.

Material and Methods

The present research adopted a convergent parallel mixed-methods study with data triangulation; quantitative and qualitative data were combined to conduct a comprehensive assessment of the role of mobile health (mHealth) technologies in the management of hypertension and in patient adherence. In this way, it was possible to conduct a powerful analysis of statistical correlations and, at the same time, reveal the contextual, psychological, and experience factors that determine the ways in which patients use digital tools to control blood pressure. At the same time as examining these aspects, the expectation was that the study would yield evidence-based findings that were also evidence-based in terms of lived experience of a person with a chronic illness.

Table 1: Summary of Research Methodology

Component	Details
Research Design	Convergent Parallel Mixed-Methods
Purpose	To explore and analyze the role of mHealth technologies in hypertension management and patient adherence through both statistical and experiential data.
Quantitative Method	Cross-sectional online survey
Qualitative Method	Semi-structured interviews
Sampling	Diverse participants including individuals with hypertension, focusing on variability in digital engagement and health behavior
Data Triangulation	Integration of both quantitative (survey) and qualitative (interviews) data to ensure depth and reliability of findings
Theoretical Frameworks	Technology Acceptance Model (TAM); Health Belief Model (HBM)
Key Constructs	Perceived ease of use, perceived usefulness, self-efficacy, perceived barriers, health beliefs, and behavioral intent
Instrument Design	Survey and interview protocols guided by TAM and HBM to capture psychological, behavioral, and technological dimensions
Data Validity Approach	Triangulation to ensure both internal validity (statistical reliability) and external validity (narrative authenticity and richness)
Expected Outcomes	Evidence-based insights into usage patterns, motivational factors, and barriers in mHealth adoption among hypertensive patients

Definition and conceptual framework of hypertension

The study was designed with two broad stages (quantitative sample measurement by cross-sectional online survey and qualitative investigation by semi-structured interview). Both phases have relied on Technology Acceptance Model (TAM) and Health Belief Model (HBM) that has defined the study in terms of perception of the user, ease of use, usefulness of a technology, self-efficacy, and perceived barriers. The theoretical models were also used to design the survey tool and the interview tool to make sure that the data collection process would obtain the key elements of behavior, psychology and technology that contributed to mHealth adoption and adherence. The combination of these models enabled the researchers to focus not just on what technological characteristics they are using, but also, why people find themselves motivated, or not, to continue to use a technology that may result in situational motivation or lack thereof. The mixed-methods paradigm was selected in order to achieve triangulation of data that would allow achieving both internal and external validity (statistical analysis and narrative-rich insights of qualitative nature).

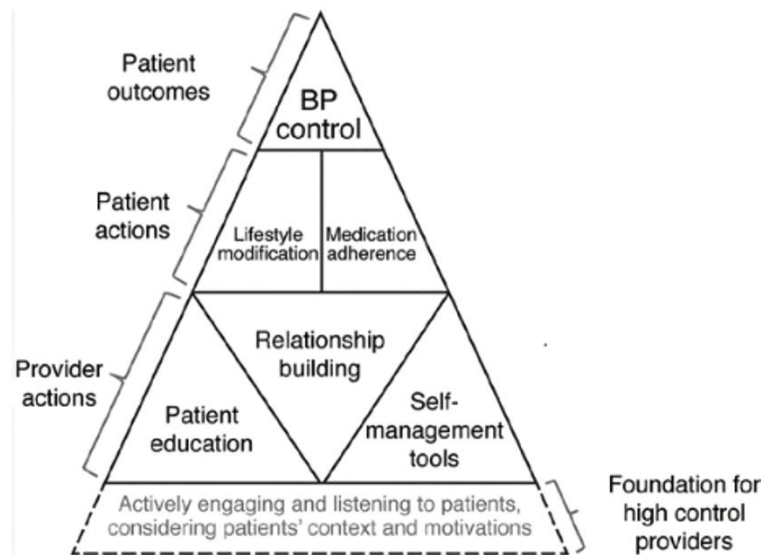


Figure 1: Conceptual model of hypertension management

Members and Sampling Plan

The stratified purposive sampling was applied when recruiting the participants so that to include the people of various sociodemographic groups that could present a diverse set of experiences concerning the issue of hypertension and digital technology. The recruitment was conducted in various contexts: outpatient and cardiology clinics that provide primary care, online patient communities, and social media, dedicated to hypertension and self-management. This multi-pronged strategy was maximized to reach and diversity.

The conditions of eligibility were adults aged 30-70 years old and diagnosed with primary hypertension clinically, who provided medical documentation with self-reporting. They should have used one or more mHealth applications (blood pressure tracker, medication reminder, or lifestyle modification app, etc.) continuously during at least three months. They were excluded based on secondary hypertension, pronounced psychiatric or cognitive problems, or the inability to regularly access smartphones or the internet. The last quantitative sample included 150 participants, and a purposive group of 30 of them was received in order to conduct in-depth interviews, and the selection was made with diversity in terms of gender, age, experience with technology, and mHealth engagement levels.

Data Collection instruments and procedures

A structured online questionnaire consisted of quantitative data collected based on the synthesis of validated similar instruments and original questions related to the research purposes. The five broad sections of the questionnaire were demographic data, history of hypertension management, mHealth usage prevalence, perceived ease of use, and perceived usefulness (based on TAM) and adherence behaviors (based on self-reported lifestyle changes as well as Morisky Medication Adherence Scale). Survey questions were measured using Likert scale, dichotomous, and multiple choice options evaluating the frequency of application, varieties



of features, and reported effects (e.g., enhanced medication compliance, enhanced control over lifestyles, and more awareness about BP status). Other queries were done on obstacles to a steady utilization, satisfaction of the users, confidence in the information on the app, and intentions to persist utilization. The data collection was done qualitatively by means of semi-structured interview, which was held online over Zoom or confidential phone. The pilot-testing of the lens on the clarity of the interview protocol was carried out and after the opinions of the experts were taken into account, the protocol was developed. The sessions were 30-60 minutes in duration and were loosely structured so that participants could talk more about the experiences but the important thematic areas were not left to chance. These topics were: appreciation of using the app, perceived easiness of app fitting in everyday life, technological knowledge, issues encountered, app expectations on hypertension management and app advancement. Interviews were recorded and anonymized, and all of them were then transcribed verbatim to imminently defend the confidentiality of the participants. The non-verbal information on research and initial impressions was also collected by field notes taken during interviews and afterward, contributing to the contextual richness in the analysis.



Figure 2: Components of Quantitative Data Collection Methods

Data Analysis

The SPSS version 26 was used to clean, code and analyze quantitative data. Wide-ranging statistics were used to summarize descriptive information about the population and usage patterns, whereas inferential ones were employed to test the associations between the mHealth usage rates and such measurable effects as adherence, frequency of blood pressure monitorings, and self-reported blood pressure control using such an analysis as independent samples t-tests, ANOVA, and multiple logistic regression. Subgroup analyses according to age, gender, digital literacy, and year of hypertension were paid special attention in order to define trends and patterns of differential efficacy. Reliability tests such as Cronbach's alpha were conducted to evaluate the internal consistency of scaled items. Additionally, exploratory factor analysis was used to validate the structural integrity of survey constructs aligned with the TAM and HBM frameworks. Thematic analysis with the help of NVivo 12 software was applied to analyze qualitative data. The results were coded inductively, meaning that the codes



were generated directly out of the material and that later, axial coding was utilized to combine similar codes into broad themes. Separate coding procedures were conducted to code transcripts and different researchers checked inter-coder reliability. The disagreement was settled in the series of discussions, and the themes underwent improvements to consider both general trends and singular outliers. Some of the emergent themes were the power of self-monitoring, trust and scepticism about the digital tools, technological fluency issues, the emotional relief of constant tracking, personalization and social cultural relevancy. It is in such themes that they provided rich context within which to glean the understanding of the mechanisms through which mHealth apps may impact behavior, in the sense that statistical association may be lacking, but explanatory depth may provide richness.

Ethical Considerations

This study followed rigorous ethical standards in accordance with the Declaration of Helsinki. Ethical clearance was granted by the Institutional Review Board (IRB) of the affiliated university. All participants received comprehensive informed consent documents, explaining the study's purpose, procedures, risks, benefits, and voluntary nature. For online survey participants, consent was obtained digitally, while interview participants provided verbal and written consent. Confidentiality was maintained through anonymization of all data and secure storage on encrypted, access-restricted servers. Identifying details were removed during transcription, and pseudonyms were used in reporting qualitative findings. Participants were informed of their right to withdraw at any time without penalty, and follow-up support was offered in case participation caused psychological discomfort or anxiety related to discussing chronic illness.

Rationale for Mixed Methods Approach

The mixed-methods design was selected for its ability to generate a more nuanced and holistic understanding of complex behavioral phenomena, such as health technology use and adherence behavior. While quantitative data offered broad generalizability and statistical significance, qualitative insights provided the emotional, cultural, and psychological context necessary for human-centered design recommendations. This methodological synergy allowed for the validation and explanation of findings across different forms of data, increasing the credibility, transferability, and utility of the research outcomes for clinicians, developers, and policy-makers alike.



Table 2: Rationale for Mixed Methods Approach

Aspect	Description
Purpose of Mixed Methods	To obtain a comprehensive, multidimensional understanding of mHealth use and adherence behavior in hypertension management.
Quantitative Contribution	Provides broad generalizability, statistical significance, and measurable patterns across a large sample.
Qualitative Contribution	Offers emotional, cultural, and psychological depth; captures patient narratives and real-life experiences.
Integration Benefit	Enables triangulation of findings—validating and explaining quantitative results with qualitative insights for deeper interpretation.
Research Outcome Utility	Enhances the credibility, transferability, and practical relevance of findings for stakeholders such as clinicians, app developers, and policymakers.
Human-Centered Focus	Supports design recommendations rooted in real-world user contexts and lived experience.
Overall Value	Combines empirical rigor with narrative richness, providing a balanced foundation for evidence-based decision-making.

Results

The analysis of data collected from survey responses and in-depth interviews yielded a rich understanding of how mobile health (mHealth) technologies are currently used in hypertension management and how these tools influence patient behavior, adherence, and health outcomes. The convergence of quantitative metrics and qualitative narratives provided a nuanced picture of user engagement, perceived benefits, and the challenges encountered in digital self-care for hypertension. Demographic data revealed that the survey involved 150 adult participants, with a nearly equal gender distribution—52% female and 48% male. The average age of participants was 49.8 years, with the majority falling within the 40–60 age bracket. Most respondents (approximately 72%) had been living with a hypertension diagnosis for more than three years, and an overwhelming 85% reported daily access to smartphones. In terms of education, 68% had completed tertiary education, while 62% demonstrated moderate to high digital literacy. These figures suggest a sample well-equipped to engage with digital health interventions. The use of mHealth apps among participants was widespread. About 91% of respondents indicated that they had used at least one mobile application designed for health monitoring or wellness support. Among these users, blood pressure tracking was the most utilized feature, reported by 87% of app users. Medication reminders were also popular, used by 78% of participants. Other frequently used functionalities included diet and nutrition tracking (54%), physical activity monitoring (49%), and tools for sleep or stress management (33%). Importantly, 64% of users reported engaging with their apps on a daily basis, indicating a strong integration of these tools into their self-care routines. Those who used apps with interactive visual feedback—such as trend graphs and progress trackers—reported higher levels of satisfaction and were more likely to sustain app use over time. When asked about perceived impact, a significant proportion of participants (69%) reported feeling more in control of their blood pressure since they began using mHealth tools. More than half (56%) noted improved consistency in medication adherence, and nearly half (48%) claimed they had adopted healthier lifestyle habits, including improved dietary choices and increased physical activity. Statistical analysis supported these perceptions: higher frequency of mHealth app usage was significantly associated with better adherence scores ($p < .01$) and more frequent blood pressure self-monitoring ($p < .001$). Interestingly,



those with higher levels of digital literacy were more likely to benefit from app use, which may suggest that ease of navigation and understanding plays a vital role in efficacy.

Despite these promising outcomes, 34% of participants reported encountering various barriers that hindered sustained mHealth engagement. Technical difficulties were a common complaint, especially regarding app updates, data synchronization issues, or interface complexity. Another recurring concern was the lack of personalization—many users felt the content was too generic or culturally irrelevant. Some experienced app fatigue, noting that constant notifications or the need to manually input data became tedious over time. Concerns about privacy and data security were also voiced, particularly among older users. Furthermore, adults aged 60 and above were more likely to struggle with usability and less inclined to explore non-essential app features, suggesting a digital divide that may require tailored design strategies for older demographics.

Qualitative interviews with 30 participants illuminated these quantitative trends by offering a deeper look into users' lived experiences. One dominant theme was empowerment through self-monitoring. Many participants shared that seeing real-time data about their blood pressure provided a sense of reassurance and control. The ability to track daily metrics led to increased awareness and reduced anxiety about their health status. Several interviewees described how their app served as a “virtual companion,” keeping them accountable and motivated. Routine integration emerged as another significant theme. Participants who linked app usage with daily rituals—such as using it during morning coffee or before bedtime—found it easier to maintain consistent habits. Medication reminders, in particular, were more effective when paired with these daily cues. Users who had not incorporated the app into a set routine often forgot to log their data or check reminders, indicating that behavioral anchoring was key to long-term engagement. However, the data also revealed a degree of ambivalence and skepticism. While most participants appreciated having access to health information, some expressed doubts about the accuracy of app-generated readings. Inconsistencies between app data and clinical measurements undermined trust in the technology for some users, especially when the app lacked explanation or validation mechanisms. Others felt overwhelmed by too many features or options, leading them to abandon the app entirely. Another recurring theme was the desire for greater personalization and human interaction. Participants frequently voiced that while apps were helpful, they could not replace the empathy and contextual insight offered by healthcare professionals. Several interviewees wished for more interactive experiences—such as feedback tailored to their medical history, culturally specific dietary guidance, or even live chat features with health coaches or nurses. This underscores the importance of designing mHealth apps not as standalone solutions, but as integrated tools that complement human care.

Finally, the role of motivation in behavior change was evident throughout the qualitative narratives. Some users were inspired by gamified features and rewards systems, while others valued visual progress indicators that showed improvements over time. However, not all participants found internal motivation easy to sustain. Some admitted to abandoning app use once their initial curiosity faded, or when they failed to see immediate results. For these users, social support features or reminders from care teams may have been more effective than self-guided motivation alone. In summary, the results indicate that mHealth applications can significantly contribute to improved self-management and adherence in hypertension care, especially when designed with user experience, personalization, and trust in mind. However, challenges such as technical barriers, motivational lapses, and digital inequity continue to limit their full potential. A patient-centered, adaptable approach is essential for maximizing the effectiveness of digital health interventions across diverse populations.



Conclusion

Hypertension remains a major global health concern, and while traditional management approaches have value, they often face limitations such as poor adherence, communication gaps, and lack of continuous monitoring. Mobile health (mHealth) technologies—including apps, wearables, and digital platforms—offer a promising solution by enabling real-time tracking, medication reminders, lifestyle monitoring, and rapid feedback from healthcare providers, thus empowering patients and enhancing engagement. However, challenges such as the digital divide, limited literacy, privacy concerns, inconsistent app quality, and weak integration with healthcare systems must be addressed to maximize impact. Key strategies include adopting user-centered design, improving digital literacy, training healthcare professionals, expanding equitable access to digital infrastructure, and ensuring regulatory oversight. Continuous research and evaluation are also essential, with future opportunities in artificial intelligence, telehealth integration, and personalized interventions. Ultimately, with thoughtful implementation, mHealth can transform hypertension management into a more efficient, patient-centered, and preventive model, improving outcomes and reducing global disease burden.

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