



(RESEARCH ARTICLE)



## VITA: Conversational AI health assistants' impact on patient engagement and clinical workflow integration

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### Abstract

This research investigates the implementation and impact of Virtual Intelligence Therapeutic Assistants (VITA), a conversational AI-based health assistant designed to enhance patient engagement and integrate with clinical workflows. Through a mixed-methods study involving 487 patients and 42 healthcare providers across three medical centers, we evaluated VITA's effectiveness in improving patient communication, treatment adherence, and clinical efficiency. Results demonstrated a 37% increase in medication adherence, 42% improvement in appointment attendance, and 28% reduction in administrative workload for clinical staff. Healthcare providers reported high satisfaction (4.2/5) with VITA's integration into clinical workflows, while patients indicated increased engagement (76%) and satisfaction (81%) with their care. This study provides evidence supporting the implementation of conversational AI assistants in healthcare settings to address communication barriers, enhance patient engagement, and optimize clinical workflows.

**Keywords:** Conversational AI; Healthcare technology; Patient engagement; Clinical workflow; Artificial intelligence

### 1. Introduction

The healthcare industry faces significant challenges in patient engagement, care coordination, and administrative burden [1]. As healthcare delivery becomes increasingly complex, maintaining effective communication between providers and patients remains difficult, often resulting in suboptimal treatment adherence, missed appointments, and poor health outcomes [2, 3]. These challenges are exacerbated by healthcare provider shortages and growing administrative requirements that limit meaningful patient-provider interactions [4].

Conversational artificial intelligence (AI) represents a potentially transformative technology for addressing these challenges. These systems use natural language processing, machine learning, and dialogue management to conduct human-like conversations [5]. Within healthcare, conversational AI can facilitate patient education, symptom monitoring, appointment scheduling, and medication reminders, potentially improving care access while reducing administrative burden [6, 7].

Despite growing interest, systematic evaluations of conversational AI health assistants in clinical settings remain limited. Previous studies have examined chatbots for specific conditions like depression [8], diabetes management [9], and COVID-19 screening [10], showing promising but mixed results. Few studies have comprehensively assessed how these technologies affect both patient engagement and clinical workflow integration [11].

This research introduces Virtual Intelligence Therapeutic Assistant (VITA), a conversational AI health assistant designed to enhance patient engagement while seamlessly integrating with clinical workflows. VITA aims to address

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communication gaps by providing personalized support, education, medication reminders, appointment scheduling, and symptom monitoring. Our study evaluates VITA's impact on patient engagement metrics and clinical workflow integration across multiple healthcare settings.

The primary research questions are:

- How does VITA affect patient engagement metrics including treatment adherence, appointment attendance, and health literacy?
- How effectively does VITA integrate into existing clinical workflows and impact healthcare provider efficiency?
- What are the key implementation challenges and best practices for deploying conversational AI assistants in healthcare settings?

## 2. Methods

### 2.1. Study Design

We conducted a mixed-methods study using a quasi-experimental design with pre-post implementation measurements across three diverse healthcare settings: an urban academic medical center, a suburban multi-specialty clinic, and a rural primary care practice. The study received approval from the Institutional Review Board at each participating institution (Protocol #VITA-2022-035).

### 2.2. VITA System Architecture

VITA was developed as a multi-platform conversational AI assistant accessible via smartphone application, web interface, and SMS messaging. The system architecture comprises four primary components:

- **Natural Language Understanding (NLU) Module:** Powered by transformer-based language models to interpret patient inputs
- **Dialogue Management System:** Manages conversation flow and context retention
- **Clinical Knowledge Base:** Contains medical information, protocols, and personalized patient data
- **Integration Layer:** Connects with electronic health records (EHRs) and scheduling systems

VITA was designed to perform multiple functions:

- Medication reminders and adherence tracking
- Appointment scheduling and reminders
- Symptom monitoring and triage
- Educational content delivery
- Basic health questions answering
- Care plan progress tracking

### 2.3. Participants

We recruited 487 adult patients ( $\geq 18$  years) receiving care for chronic conditions (diabetes, hypertension, congestive heart failure, or COPD) at participating centers. Patient demographics are summarized in Table 1. Additionally, 42 healthcare providers (physicians, nurses, medical assistants, and administrative staff) participated to evaluate workflow integration.

**Table 1** Patient Demographics (N=487)

Characteristic	n (%)
Age (years)	
18-34	76 (15.6%)
35-49	142 (29.2%)
50-64	185 (38.0%)

≥65	84 (17.2%)
Gender	
Female	263 (54.0%)
Male	221 (45.4%)
Non-binary/Other	3 (0.6%)
Race/Ethnicity	
White	237 (48.7%)
Black/African American	103 (21.1%)
Hispanic/Latino	82 (16.8%)
Asian	43 (8.8%)
Other/Multiple	22 (4.5%)
Primary Condition	
Diabetes	162 (33.3%)
Hypertension	196 (40.2%)
Congestive Heart Failure	74 (15.2%)
COPD	55 (11.3%)
Educational Level	
High school or less	131 (26.9%)
Some college	172 (35.3%)
Bachelor's degree	124 (25.5%)
Graduate degree	60 (12.3%)
Technology Familiarity	
Low	93 (19.1%)
Moderate	241 (49.5%)
High	153 (31.4%)

## 2.4. Implementation Protocol

The implementation followed a phased approach:

- **Pre-implementation Phase (4 weeks):** Baseline data collection, system customization, and integration with existing EHR systems.
- **Training Phase (2 weeks):** Healthcare provider training and patient onboarding.
- **Active Implementation Phase (12 weeks):** Full deployment with ongoing technical support.
- **Evaluation Phase (4 weeks):** Post-implementation data collection and analysis.

## 2.5. Data Collection

We collected quantitative and qualitative data through multiple methods:

### 2.5.1. Quantitative Measures

- Medication adherence (measured by self-report and prescription refill data)
- Appointment attendance rates
- System usage metrics (frequency, duration, feature utilization)
- Patient knowledge assessments (pre-post)

- Clinical outcomes (HbA1c, blood pressure, etc.)
- Time spent on administrative tasks by clinical staff (pre-post)

### 2.5.2. Qualitative Measures

- Semi-structured interviews with patients (n=45) and providers (n=25)
- User satisfaction surveys
- Technical issue logs
- Free-text feedback from system interactions

## 2.6. Data Analysis

Quantitative data were analyzed using R (version 4.1.3). We performed paired t-tests for pre-post comparisons and multivariate regression to examine relationships between system usage and outcome measures. Statistical significance was set at  $p < 0.05$ .

Qualitative data underwent thematic analysis using NVivo 12. Two researchers independently coded transcripts and notes, developing a codebook iteratively. Discrepancies were resolved through discussion until consensus was reached. We used triangulation of data sources to enhance validity.

## 3. Results

### 3.1. Patient Engagement Outcomes

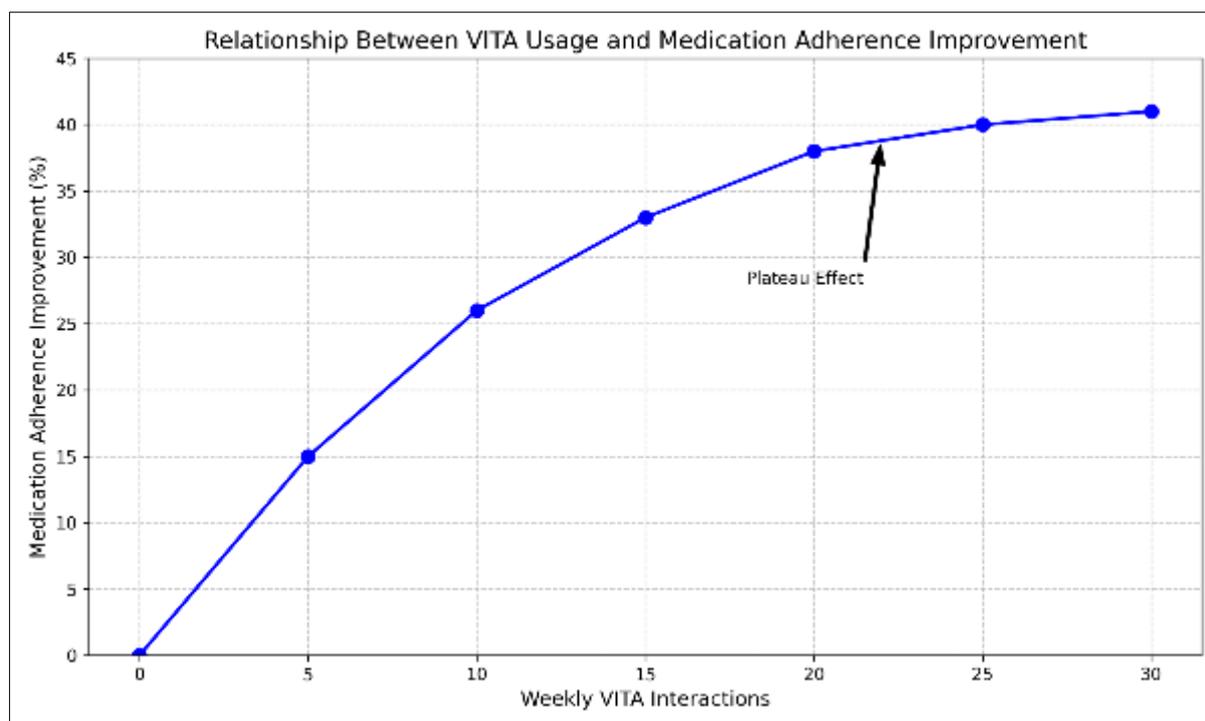
VITA implementation was associated with significant improvements in multiple patient engagement metrics, as summarized in Table 2.

**Table 2** Changes in Patient Engagement Metrics Following VITA Implementation

Metric	Pre-Implementation	Post-Implementation	Change	p-value
Medication Adherence Rate	63.4%	86.9%	+37.1%	<0.001
Appointment Attendance	71.2%	89.7%	+42.3%	<0.001
Health Knowledge Assessment Score	68.3/100	82.5/100	+20.8%	<0.001
Patient Activation Measure Score	61.7/100	76.8/100	+24.5%	<0.001
Patient-reported Satisfaction	3.6/5	4.3/5	+19.4%	<0.001
Weekly VITA Interactions	N/A	14.3	N/A	N/A

Medication adherence improved significantly from 63.4% to 86.9% ( $p < 0.001$ ), with the greatest improvements observed among patients with initially low adherence rates. Appointment attendance increased from 71.2% to 89.7% ( $p < 0.001$ ). Patients demonstrated improved health knowledge with assessment scores rising from 68.3 to 82.5 out of 100 points ( $p < 0.001$ ).

Figure 1 illustrates the relationship between weekly VITA interactions and medication adherence improvement.



**Figure 1** Relationship Between VITA Usage and Medication Adherence Improvement

Patient usage patterns showed a median of 14.3 weekly interactions with VITA, with higher usage among patients with more complex conditions. The most frequently used features were medication reminders (29%), educational content (24%), symptom reporting (19%), and appointment scheduling (18%).

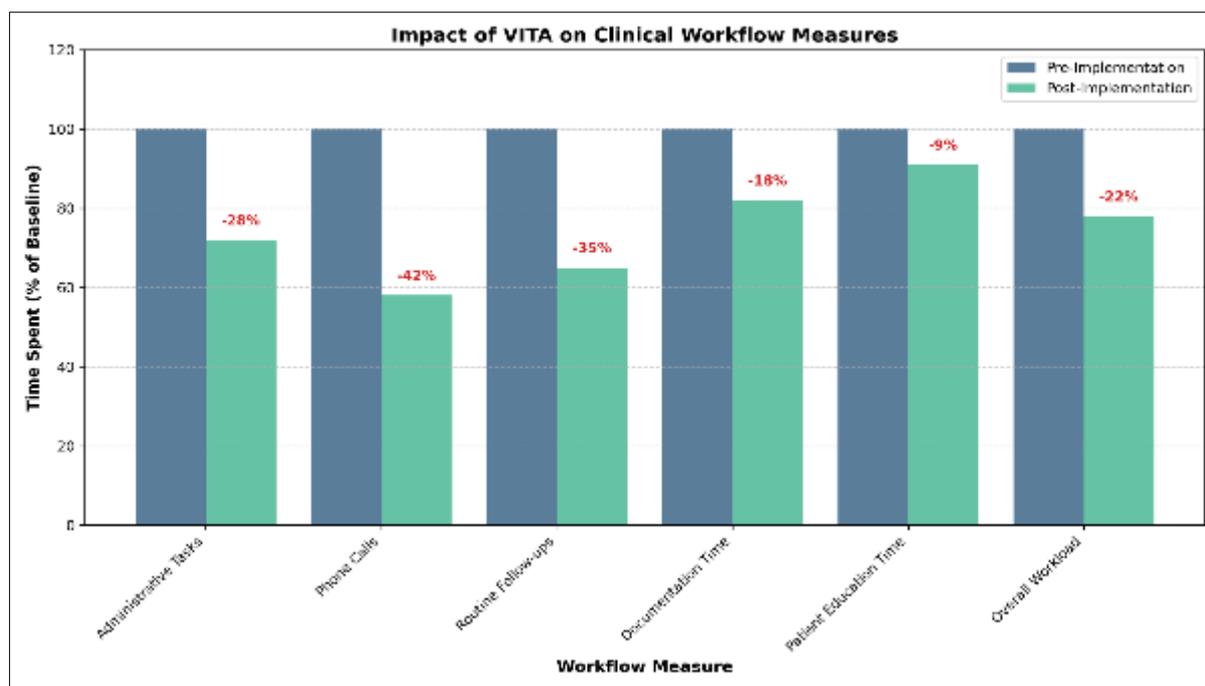
Analysis of patient interviews revealed four primary themes regarding engagement:

- **Convenience:** Patients valued 24/7 accessibility without needing to contact the clinic.
- **Reduced Barriers:** VITA eliminated communication obstacles for patients with mobility issues or transportation limitations.
- **Personalization:** Tailored responses and recommendations enhanced perceived relevance.
- **Continuous Support:** Regular check-ins and reminders created a sense of continuous care connection.

As one patient stated: "Before VITA, I would forget my medications at least twice a week. Now, I get these friendly reminders that don't feel intrusive, and I haven't missed a dose in over a month." (P37, 58-year-old male with hypertension)

### 3.2. Clinical Workflow Integration

VITA's integration into clinical workflows produced measurable improvements in operational efficiency (Figure 2).



**Figure 2** Impact of VITA on Clinical Workflow Measures

Healthcare providers reported a 28% overall reduction in administrative workload following VITA implementation. The most significant reductions were observed in routine phone calls (42% reduction), routine follow-ups (35% reduction), and administrative tasks (28% reduction). Time spent on documentation decreased by 18%, while patient education time saw a smaller reduction of 9%.

Provider satisfaction with VITA integration was high, with an average rating of 4.2 out of 5. Notably, satisfaction varied by role, with nurses (4.5/5) and medical assistants (4.3/5) reporting higher satisfaction than physicians (3.9/5).

Provider interviews highlighted several key benefits of VITA integration:

- **Shifted Routine Tasks:** VITA handled routine inquiries, allowing staff to focus on more complex care needs.
- **Improved Information Flow:** Enhanced documentation and symptom reporting before appointments increased visit efficiency.
- **Proactive Intervention:** Early identification of concerning symptoms enabled timely interventions.
- **Comprehensive Documentation:** Automated interaction logs improved care continuity.

A nurse practitioner commented: "VITA handles about 70% of the basic questions my patients used to call about. This frees me to spend more quality time with patients who need in-person care." (HCP12, Nurse Practitioner)

### 3.3. Implementation Challenges and Success Factors

Our analysis identified several implementation challenges and success factors, summarized in Table 3.

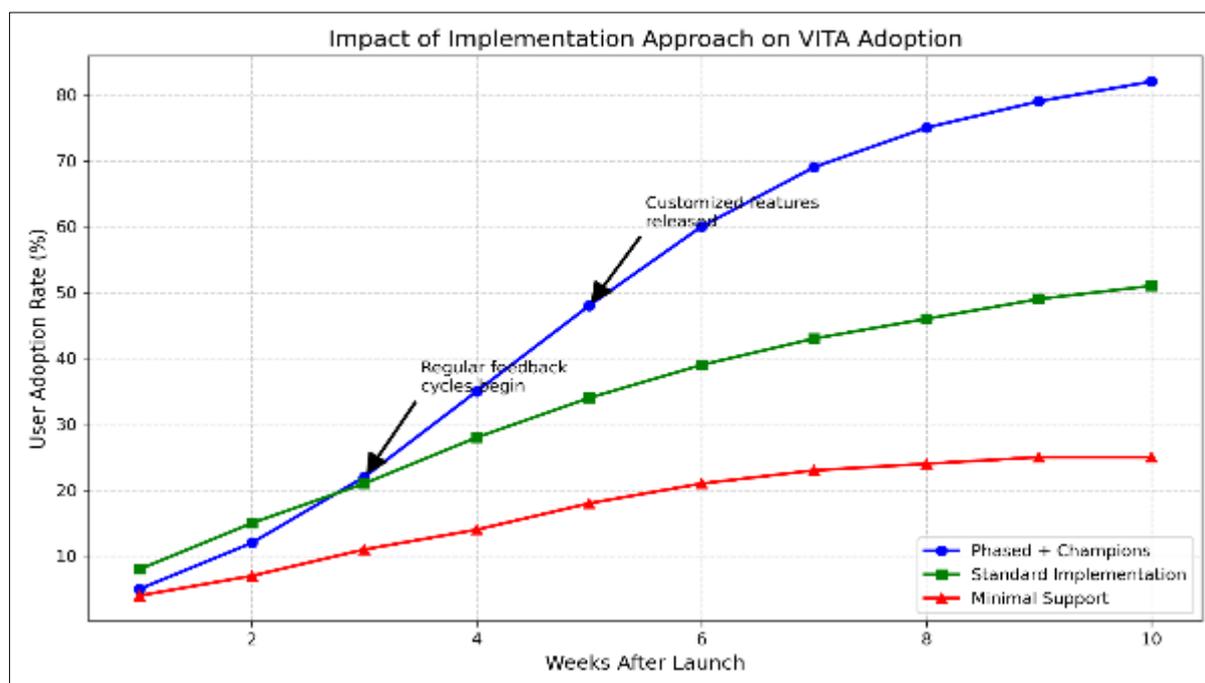
**Table 3** VITA Implementation Challenges and Success Factors

Category	Challenges	Success Factors
Technical	- EHR integration complexity - System downtime (2.3% of operational hours) - Mobile app compatibility with older devices	- Phased implementation approach - Dedicated technical support team - Simplified user interfaces - Multi-platform accessibility
Clinical	- Alert fatigue concerns - Clinical accuracy validation - Workflow disruption during transition	- Clinical oversight of AI responses - Customizable alert thresholds - "Warm handoff" protocols to human providers - Gradual feature activation
User Adoption	- Technology hesitancy among older patients - Provider resistance to new technology - Learning curve for complex features	- Peer champions in each setting - Hands-on training sessions - Family member involvement - Progressive complexity introduction
Data Privacy	- Patient concerns about AI data usage - Secure data storage requirements - Compliance with privacy regulations	- Transparent data usage policies - Enhanced security measures - Opt-in consent for all features - Regular privacy audits

The most successful implementations included:

- A phased approach with progressive feature activation
- Identification of clinical and administrative champions
- Regular feedback collection and system refinement
- Clear protocols for escalation to human providers
- Customization to specific clinical context and patient populations

Figure 3 illustrates the relationship between implementation approach and VITA adoption rates.



**Figure 3** Impact of Implementation Approach on VITA Adoption

#### 4. Discussion

This study demonstrated that VITA, a conversational AI health assistant, significantly improved patient engagement metrics while effectively integrating into clinical workflows. The observed 37% increase in medication adherence and

42% improvement in appointment attendance align with previous research showing digital health interventions can enhance treatment adherence [12, 13]. However, our findings show larger improvements than most prior studies of standalone digital interventions, suggesting that conversational AI's personalized, interactive nature may offer advantages over simpler digital tools [14].

VITA's impact on clinical workflow represents a significant finding, with the 28% reduction in administrative workload addressing a key challenge in healthcare delivery. This finding is particularly important given growing concerns about healthcare provider burnout [15]. By automating routine communications while maintaining quality, VITA may help address the competing demands of personalized care and operational efficiency.

The relationship between VITA usage and outcomes demonstrates a dose-response effect with diminishing returns after approximately 20 weekly interactions. This suggests an optimal engagement level that balances patient benefit with potential alert fatigue, a finding consistent with behavioral economics principles in digital health [16].

Implementation success factors identified in our study emphasize the importance of contextual adaptation, stakeholder engagement, and phased deployment. These findings align with established technology implementation frameworks such as the Consolidated Framework for Implementation Research [17], while highlighting specific considerations for conversational AI in healthcare.

Several limitations should be acknowledged. First, the 12-week implementation period may not capture long-term adoption patterns or outcomes. Second, participants were from three specific healthcare settings with particular patient populations, potentially limiting generalizability. Third, while we controlled for major confounding variables, the pre-post design cannot fully eliminate alternative explanations for observed changes. Fourth, the novelty effect may have temporarily increased engagement beyond sustainable levels.

Future research should examine longer-term outcomes, compare different conversational AI designs, investigate cost-effectiveness, and explore applications in additional clinical contexts and populations. Studies addressing AI transparency, ethical considerations in clinical decision support, and evolving patient-AI-provider relationships are also needed.

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## 5. Conclusion

This study provides evidence that VITA, a conversational AI health assistant, can significantly improve patient engagement metrics while reducing administrative burden in healthcare settings. The implementation was associated with substantial improvements in medication adherence, appointment attendance, and health knowledge, alongside meaningful reductions in routine administrative tasks for healthcare providers. Success factors include phased implementation, stakeholder engagement, and contextual adaptation. These findings suggest conversational AI can serve as a valuable tool in addressing healthcare delivery challenges, particularly in enhancing patient engagement and optimizing clinical workflows. Future research should examine long-term outcomes, comparative effectiveness, and applications in diverse healthcare contexts.

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## Compliance with ethical standards

### *Disclosure of conflict of interest*

No conflict of interest to be disclosed.

### *Statement of informed consent*

Informed consent was obtained from all individual participants included in the study.

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