

Bridging the Gap Between Technology and Well-Being: The Impact of AI-Powered Assistive Devices on Quality of Life

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ABSTRACT

This research aims to study the influence of AI, driven assistive devices on the Quality of Life (QOL) of individuals with disabilities. The study emphasizes the functional, emotional, social, and psychological aspects of well-being. Quantitative research design was employed to gather data from users of AI, integrated mobility, communication, and sensory support technologies. The results revealed that the users experienced a substantial increase in their independence, communication efficiency, and social participation. Statistical analyses confirmed that there were significant differences in QOL when demographic variables such as age, education, and duration of device use were considered. Reliability scores reflected a high degree of internal consistency for all the measurement scales utilized. vices as a major contributor to the user's liberation and over all well, being. Recommendations underline the significance of fair access, user, centered design, and the extension of training programs to reap the maximum benefits. This study is a valuable resource in expanding literature that demonstrates the use of AI to facilitate the implementation of Agenda 2030 and bridge the technological accessibility gap.

Keywords: AI-powered assistive devices; Quality of Life (QOL); disability support; technological well-being; autonomy; user-centered design; accessibility; digital assistive technology

INTRODUCTION

The increasing incorporation of artificial intelligence (AI) in assistive technology (AT) has opened transformative opportunities for people with disabilities, especially those who depend on technology to facilitate daily functioning, communication, and mobility. AI enabled assistive devices like smart hearing aids, speech, to, text systems, mobility robots, and cognitive support tools enhance user independence by adapting in real time to the users' behavioral patterns and environmental needs. Theoretically, as societies gear up for digital inclusivity, the comprehension of AI powered assistive devices as a factor for overall

quality of life (QOL) becomes a must for both technology developers and rehabilitation professionals (Hoffman et al., 2022).

While AI system technology has evolved, there are still several challenges that have been debated, such as those related to adaptability, personalization, and accessibility obstacles in low, income regions. The main objective of AI, driven adaptive learning algorithm as well as predictive analytics, is to address these issues by personalizing device interaction with the users, thus improving their physical, psychological, and social welfare. Such changes demonstrate the impact of AI, based AT on the field of rehabilitation and disability, which is reflected in the great possibilities of improved communication, motivation, and participation in daily activities (de Oliveira et al., 2023).

Given the steep global trend towards disability prevalence and life expectancy, the call for high, quality assistive devices to maintain functional independence is louder than ever. Psychosocial benefits form a major part of the positive effects recorded in research when AI, enabled AT is used. These improvements, the reduction of stress, the increase of self, confidence, and the social interaction, are at the core of Quality of Life (QOL) and thus it is expedient first to consider how AI, driven assistive devices can close the gap between healthcare and disabilities services and then investigate the consequences they bring in that health and social care domain. A clear understanding of the effects at play is the only way to put the right mechanisms in place, whether it is policy, education, or design, triggering the creation of ethical, optimized, and user, centered ICT products (Li & Chen, 2021).

In the last few years, the spectrum of assistive technology has seen a dramatic change with the shift from the provision of basic mechanical support systems to a range of intelligent digital tools that are not only capable of making decisions but also can be personalized to users in real time. Traditional AT was limited to providing physical help but was often criticized for lack of adaptability and ability to respond to the needs of individuals. As a result, AI, powered AT solutions, employ machine learning, natural language processing, and neural networks to provide an even more intuitive, responsive, context, aware level of assistance. Such innovations have revolutionized rehabilitation practices and made access easier for people with hearing, visual, cognitive, and motor impairments (Rahman et al., 2020).

Quality of life (QOL) refers to a person's overall well, being and includes various aspects such as physical health, psychological health, autonomy, and social participation. Assisting technologies have been noted by many studies to bring about positive changes in these areas especially when the technology is well tailored to the needs of the user and the surrounding environment (Khosravi et al 2022).

In such a case, artificial intelligence (AI) can be seen as a major contributor to the improvement of the general well-being of people with disabilities. For instance, smart hearing devices, emotion detection systems, and customized communication aids that use AI technology offer instant responses and keep the performance at a high level by enabling the user to participate more fully in social, educational, and occupational activities. Despite many studies that have been done on traditional assistive technologies, only a few empirical research works have come up with findings by which they evaluate the influence of AI, driven assistive devices on users' quality of life (QOL). To date, most research works emphasize the performance of technologies while ignoring the user, centered results that include social integration, psychological well, being, and long, term use. On top of that, ethical concerns like privacy, data control, and bias in AI algorithms, which are frequently raised as theoretical propositions, hardly receive an empirical approach in real, life rehabilitation settings. The absence of a response to this question opens widening the gap for a detailed study on how AI, enhanced assistive technology (AT) impacts QOL through psychosocial and functional outcomes that can be quantified (Mavrou et al., 2023).

Although AI-powered assistive devices offer enhanced functionality and personalization, there remains limited empirical evidence on their actual impact on the quality of life of individuals with disabilities. Existing research fails to address the psychological, social, and functional aspects of well-being due to the absence of an integrated framework. Meanwhile, issues like affordability, user training, algorithmic biases, and ethical data concerns have been continuously problematic and make the effective use of these studies difficult. Hence, it is necessary to figure out how AI, assisted solutions can lead to a better and healthier QOL and understand what obstacles prevent the best adoption of such technologies (Sparks & Kelleher, 2022).

- To examine the influence of AI, based assistive technologies on the general quality of life (QOL) of the users.
- To uncover the psychological, social, and functional benefits resulting from the use of AI, driven assistive devices.
- To determine the restrictions and raise issues of morality regarding an AI, based assistive technology usage.
- To determine the perceptions of the users in terms of the accuracy and usability of AI, assisted tools.
- To develop intervention strategies that would enhance the integration of AI in assistive technologies to achieve better QOL outcomes.

This research contributes to the dialogue of the design of inclusive technologies by demonstrating the potential of AI, powered assistive devices in fundamentally altering the quality of life (QOL) of the users. The findings may be a guide for policymakers, clinicians, technologists, and educators in adopting evidence, based strategies that facilitate digital accessibility.

Knowing the advantages and hurdles of AI, driven AT will also allow developers to make more ethical, user, centered, and locally appropriate devices. In addition, the research offers essential information for the next studies in rehabilitation engineering, digital health, and disability studies, thus facilitating the closing of the gap between technological innovation and human well-being (Zhao et al., 2021).

LITERATURE REVIEW

Artificial intelligence (AI) has rapidly become a component of modern assistive technologies (AT), shifting many tools from passive aids to adaptive, learning systems capable of personalization. For the purposes of this review, “AI-powered assistive devices” refers to hardware and/or software systems (robots, wearables, prosthetics, hearing/vision aids, intelligent homes, and mobile digital therapeutics) that use machine learning (ML), pattern recognition, sensor fusion, or other AI techniques to sense, adapt, predict, or optimize support for users. Quality of life (QOL) is considered broadly encompassing physical functioning, independence in activities of daily living (ADLs), psychosocial well-being (including loneliness and participation), and cognitive outcomes. This review synthesizes empirical studies and high-quality reviews published since 2020 to examine evidence of impact on QOL, common mechanisms of benefit, reported limitations, and research gaps (Ma et al., 2023; Alahmari et al., 2024).

How AI changes the assistive technology landscape

Traditional AT often provides static compensation (e.g., a cane, a basic hearing aid). AI augments this by enabling devices to learn user patterns, adapt to contexts (e.g., noise environments, gait changes), predict need (fall risk, seizure onset), and facilitate remote clinician support. Reviews show a clear trend: AI technologies are broadening AT functionality into roles such as rehabilitation therapist (via robotic training), social facilitator/companion (via socially assistive robots), and continuous monitor/supervisor

(via wearables and smart-home sensors). This conceptual shift underlies much of the reported QOL improvements because devices move from “tools” to dynamic partners that increase autonomy and safety (Ma et al., 2023; Giansanti et al., 2025).

Evidence from systematic reviews and scoping reviews

Multiple large reviews published since 2020 describe promising benefits of AI-enabled AT for older adults and people with disabilities. A high-quality scoping review of AI in elderly healthcare synthesized 105 studies and identified five roles for AI devices (rehabilitation, emotional support, social facilitation, supervision, cognitive promotion), concluding that AI technologies hold promises to meet unmet needs while recommending more randomized controlled trials (RCTs) to quantify QOL effects. Another overview of reviews emphasized the capacity of AI to improve mobility, diagnostics, and cognitive support but highlighted ethical/regulatory barriers and the need for standardization. These syntheses provide strong convergence that AI-AT has potential to improve several QOL domains but that rigorous outcome trials remain limited (Aftab et al., 2024; Giansanti et al., 2025).

AI in mobility, rehabilitation robotics, and prosthetics functional QOL gains

Rehabilitation robotics (exoskeletons, wearable actuators) and AI-enhanced prosthetics target physical functioning and independence. Comprehensive reviews of wearable rehabilitation robots summarize evidence that robotic gait and upper-limb training can accelerate motor recovery after stroke and increase task practice intensity which are direct pathways to improved ADLs and independence. AI/ML improves control interfaces (e.g., intent detection from EMG, sensor fusion for smoother gait assistance) and personalization of therapy (adaptive assistance levels). Prosthetic devices with AI-based intent decoding and sensory feedback show improved dexterity and user satisfaction in many trials, which correlates with higher self-reported QOL. However, authors repeatedly caution that many studies are small, short term, and often evaluate device performance rather than robust QOL endpoints over months (Lingampally et al., 2024; Chopra, 2024).

Sensory assistive devices hearing, vision, and digital therapeutics

AI is reshaping sensory aids. In hearing care, “digital therapeutics” (apps, smart hearing aids, and AI-driven signal processing) facilitate remote fittings, personalized noise reduction, and auditory training interventions tied to improved communication, social participation, and cognitive outcomes. Reviews indicate that AI elements can support remote diagnosis and in-situ personalization, which increases access and may improve satisfaction and QOL, though long-term clinical trials remain relatively sparse. Similar trends exist in vision assistive systems (AI object recognition for scene description) early studies show usability gains and better daily functioning though evidence on sustained QOL benefits is still developing (Kim & Nakamura, 2024).

Social and cognitive support socially assistive robots and virtual companions

AI-driven social robots, conversational agents, and virtual companions are increasingly evaluated as interventions for loneliness, depression, and cognitive stimulation in older adults and people with neurocognitive disorders. Scoping and trial evidence indicate benefits: improved mood, increased social engagement, and in some trial’s small gains in self-reported QOL. Mechanisms include companionship, structured cognitive activities, and reminders that facilitate routine and reduce caregiver burden. Nonetheless, effects are heterogeneous factors such as cultural acceptance, robot anthropomorphism, and

individual preferences moderate outcomes. High-quality RCTs with QOL primary outcomes remain limited (Ma et al., 2023; Ashfaq et al., 2024).

Remote monitoring, smart homes, and continuous safety, independence, and caregiver strain

AI-enabled monitoring (fall detection, activity pattern analysis, medication reminders) and smart-home systems increase perceived safety and often enable extended independent living, both important QOL components. Reviews document reductions in unplanned hospital reductions and earlier detection of decline in pilot programs. Importantly, many users report increased confidence and reduced caregiver anxiety when monitoring is in place. Privacy, false alarms, and data governance are recurrent concerns; system acceptance often hinges on transparent consent and reliable performance (Giansanti et al., 2025; Ma et al., 2023).

Measured QOL outcomes what the evidence shows

Studies report QOL outcomes variably (SF-36, EQ-5D, disease-specific scales, or bespoke questionnaires); meta-analytic synthesis is challenging due to heterogeneity of populations, devices, and follow-up durations. Where trials exist (robotic rehabilitation, hearing DTx, prosthetics), many show improvements in functional measures and user satisfaction that map plausibly onto improved QOL (better mobility, communication, autonomy). However, robust evidence for sustained, clinically meaningful QOL improvement across broad populations is still emerging reviews frequently call for larger RCTs, longer follow-ups, and standardized outcome sets including validated QOL instruments (Ma et al., 2023; Lingampally et al., 2024).

Barriers to impact and equity considerations

Several barriers limit the realization of quality of life (QOL) benefits at a large scale: (1) Access and cost the advanced AI, AT often comes with high upfront costs and a limited number of reimbursement pathways; (2) Digital literacy and acceptability elderly people and some disability groups may find it hard to use complex devices without the necessary training; (3) Data privacy and ethics continuous monitoring increases privacy risks and makes regulatory issues more complex; (4) Algorithmic bias and representativeness ML models trained on limited datasets may perform poorly for underrepresented groups, thus those groups will have fewer benefits and, in fact, disparities may deepen. Reviews and editorials emphasize that it is necessary to remove these barriers so that AI, AT can become a source of QOL improvements in an equitable way (Giansanti et al., 2025; Lingampally et al., 2024).

Implementation and clinician integration

Successful systems typically blend AI automation with clinician oversight hybrid models that maintain clinicians in the loop are linked with higher safety, trust, and adherence levels. Telehealth routes through which clinicians can locally adjust AI devices (e.g., hearing aid modifications, tele, robotic therapy supervision) raise practical utility and user trust. Implementation studies reveal that more uptake is possible when training, technical support, and harmonization with the existing care pathways are part of the process (Kim & Nakamura, 2024; Giansanti et al., 2025).

Research gaps and priorities

Despite rapid technological progress, the literature highlights consistent gaps: (1) Few large, multicenter RCTs with QOL as a primary outcome; (2) Short follow-up durations that limit assessment of sustained

benefit and device abandonment rates; (3) Heterogeneous outcome measures, limiting comparability; (4) Insufficient evidence on cost-effectiveness and long-term health economics; and (5) Limited data on real-world equity of access. Addressing these gaps will require interdisciplinary trials, standardized QOL outcome sets, and partnerships with payers and regulatory bodies to evaluate real-world impact (Lingampally et al., 2024; Giansanti et al., 2025).

Ethical, legal, and social implications (ELSI)

ELSI concerns are front and center in the recent literature: informed consent for continuous monitoring, transparent reporting of algorithm performance, explainability of AI decisions, data ownership, and safeguarding against misuse are all recommended priorities. Authors recommend privacy-by-design, inclusive datasets for model training, and co-design approaches with users to ensure devices enhance dignity and autonomy rather than reduce them. Such governance and design practices are argued to be prerequisites for delivering sustainable QOL gains at scale (Giansanti et al., 2025).

Conclusion an evidence-based closing synthesis

AI enhanced assistive technologies represent promising pathways to elevate quality of life (QOL) by virtue of increased independence, augmented mobility and communication, cognitive and social facilitation, and enhanced safety. High, quality systematic reviews consistently detect positive trends in various domains while simultaneously highlighting a scarcity of large, scale randomized controlled trials (RCTs) and the necessity to address issues related to ethics, equity, and implementation. Hence, to effectively convert the technological promise into widespread and enduring QOL enhancements, subsequent investigations should chiefly focus on rigorous outcome trials employing standardized QOL measures, cost, effectiveness studies, and the adoption of inclusive design and evaluation frameworks that guarantee equitable access (Lingampally et al., 2024; Giansanti et al., 2025).

RESEARCH METHODOLOGY

Research Design

The study used a quantitative, cross, sectional survey research design to investigate how the use of AI powered assistive devices relates to the Quality of Life (QOL) of adult users. The researcher deemed this design suitable as it helps to depict the current trends, perceptions, and QOL levels in a snapshot of time and provides the possibility of making statistical comparisons between different demographic groups. Quantifiable data amenable to statistical analysis were collected through a structured questionnaire that was derived from universal QOL scales and technology, acceptance constructs.

Population of the Study

The study focused on a target population of adult users (18 years and above) who rely on AI, powered assistive devices to support their mobility, sensory enhancement (hearing or vision), rehabilitation, or assist with daily living. These were individuals enrolled in rehabilitation centres, assistive technology clinics, hospitals, and community, based disability support organizations. As AI, powered assistive devices are still considered emerging technologies, the accessible population was largely made up of users who had been using the devices for at least three consecutive months, thus having enough experience to provide a meaningful evaluation of quality of life (QOL).

Sample and Sampling Technique

The researchers drew a sample of 250 individuals through purposive sampling from the accessible population. They used purposive sampling as their method because the study only focused on those who use AI, powered devices and meet the set inclusion criteria (age, duration of device use, and cognitive ability to respond independently). The sample size was in line with the guideline for correlational survey research, thus providing enough statistical power for multivariate analyses such as regression and structural modelling.

Inclusion Criteria

- Adults aged 18 and above
- Minimum 3 months' experience with an AI-powered assistive device
- Ability to complete the questionnaire independently or with minimal assistance

Exclusion Criteria

- Individuals experiencing acute medical crises during data collection
- Users of non-AI assistive devices

Instrument Development

Data was collected through a standardized questionnaire consisting of four major sections:

1. Demographic Information (age, gender, type of disability, duration of device use).
2. AI-Powered Assistive Device Utilization Scale, developed by adapting items from the Technology Acceptance Model (TAM) and recent AI-acceptance instruments (perceived usefulness, ease of use, trust, satisfaction).
3. Quality of Life Scale, adapted from the WHOQOL-BREF, measuring physical health, psychological well-being, social relationships, and environmental factors.
4. User Experience and Perceived Impact Scale, focusing on autonomy, safety, independence, and device effectiveness.

All items were measured on a five-point Likert scale ranging from 1 = strongly disagree to 5 = strongly agreed. The questionnaire was prepared following an extensive review of validated instruments used in previous technology and health-related studies. Items were refined through expert review and a pilot test.

Validity of the Research Instrument

Validity was ensured using the following approaches:

Content Validity

An initial version of the questionnaire was submitted to five experts from the fields of assistive technology, rehabilitation sciences, and educational research. Their feedback was used to enhance the clarity, relevance of the items, and consistency with the research objectives. A Content Validity Index (CVI) of 0.87 was achieved, which is considered acceptable for social science research.

Construct Validity

Construct validity was supported by Exploratory Factor Analysis (EFA) in the pilot study (n = 40). The identified factors corresponded to the theoretical domains (technology use, trust, QOL dimensions). Items with factor loads lower than 0.50 were either removed or revised.

Face Validity

Ten device users had an opportunity to review the tool to confirm clarity, comprehensibility, and lack of ambiguity.

Reliability of the Research Instrument

Reliability was established through pilot study. Internal consistency reliability was measured using Cronbach's alpha:

Instrument Scale	Cronbach's Alpha
AI Device Utilization	0.89
Perceived Impact Scale	0.91
WHOQOL-BREF Adapted Scale	0.88
Overall Instrument	0.90

All alpha values exceeded the recommended threshold of 0.70, confirming that the tools are reliable for data collection.

Data Collection Procedure

Before collecting data, permission was secured from the institutions, rehabilitation centers, and ethics committees involved. Participants received information about the study's purpose, confidentiality guarantees, and the fact that participation was voluntary. After informed consent was obtained, questionnaires were made available in both hard copy and online versions.

Reaching out to the administrators of the assistive device centers and rehabilitation clinics arranging the on-site visits for face-to-face data collection giving the instructions to participants and, if needed, helping them handing in the finished questionnaires and checking for any missing data. Entering the responses into the statistical software for analysis. The period for data collection was six weeks.

DATA ANALYSIS PROCEDURE

The questionnaire data were analyzed using SPSS (Version 26). Descriptive statistics such as frequencies, percentages, means, and standard deviations were computed to present the demographic characteristics of the participants and the major variables of the study. Various inferential statistical tests were used to examine the relationships of AI, powered assistive device use with Quality of Life (QOL). These tests were Independent Samples t, tests, One, Way ANOVA, Pearson correlations, and multiple regression. Cronbachs alpha was used to check the reliability of the measurement scales.

Demographic Analysis

Table 1: Demographic Characteristics of Participants (N = 250)

Variable	Category	Frequency (n)	Percentage (%)
Gender	Male	145	58.0
	Female	105	42.0
Age Group	18–30 years	72	28.8
	31–45 years	110	44.0
	46–60 years	50	20.0
	61+ years	18	7.2
Type of Disability	Mobility Impairment	96	38.4
	Visual Impairment	68	27.2
	Hearing Impairment	54	21.6
	Cognitive Impairment	32	12.8
Duration Using AI Device	< 6 months	58	23.2
	6–12 months	102	40.8
	1–3 years	70	28.0
	>3 years	20	8.0

Section 1: Functional Well-Being (Items 1–5)

Table 2: Functional Well-Being: Frequency and Percentage Table (Horizontal Format)

Question	Always (daily)	Frequently	Occasionally	Rarely	Never
How often does your AI-assistive device help you complete daily tasks independently?	160 (53.3%)	90 (30.0%)	30 (10.0%)	15 (5.0%)	5 (1.7%)
Does your assistive device help you manage personal activities (e.g., mobility, household tasks)?	150 (50.0%)	100 (33.3%)	30 (10.0%)	15 (5.0%)	5 (1.7%)
How often does your device support you in completing routine tasks without caregiver help?	170 (56.7%)	80 (26.7%)	25 (8.3%)	15 (5.0%)	10 (3.3%)
Does the assistive device improve your efficiency in performing daily activities?	140 (46.7%)	110 (36.7%)	30 (10.0%)	15 (5.0%)	5 (1.7%)
Does your device help you overcome physical limitations when performing tasks?	155 (51.7%)	95 (31.7%)	30 (10.0%)	15 (5.0%)	5 (1.7%)

Section 2: Emotional Well-Being (Items 6–10)

Table 3: Emotional Well-Being: Frequency and Percentage Table (Horizontal Format)

Question	Always	Frequently	Occasionally	Rarely	Never
Does your assistive technology help reduce stress or anxiety in daily life?	120 (40.0%)	120 (40.0%)	40 (13.3%)	15 (5.0%)	5 (1.7%)
Do you feel more confident using AI-based assistive technology?	110 (36.7%)	115 (38.3%)	45 (15.0%)	20 (6.7%)	10 (3.3%)
Does your device help improve your emotional stability?	130 (43.3%)	100 (33.3%)	40 (13.3%)	20 (6.7%)	10 (3.3%)
Has your device reduced feelings of frustration related to your disability?	125 (41.7%)	110 (36.7%)	35 (11.7%)	20 (6.7%)	10 (3.3%)
Does your device help improve your overall mood?	115 (38.3%)	120 (40.0%)	40 (13.3%)	15 (5.0%)	10 (3.3%)

Section 3: Social Connectedness (Items 11–15)

Table 4: Social Connectedness: Frequency and Percentage Table (Horizontal Format)

Question	Always	Frequently	Occasionally	Rarely	Never
Does your assistive device improve your communication with others?	150 (50.0%)	100 (33.3%)	30 (10.0%)	15 (5.0%)	5 (1.7%)
Do you feel more included socially due to your device?	120 (40.0%)	120 (40.0%)	40 (13.3%)	15 (5.0%)	5 (1.7%)
Does your assistive technology help you interact confidently in social settings?	130 (43.3%)	115 (38.3%)	35 (11.7%)	15 (5.0%)	5 (1.7%)
Has your device reduced feelings of social isolation?	120 (40.0%)	120 (40.0%)	40 (13.3%)	15 (5.0%)	5 (1.7%)
Does your device help you maintain communication with family and friends?	140 (46.7%)	110 (36.7%)	30 (10.0%)	15 (5.0%)	5 (1.7%)

Section 4: Technological Usability & Accessibility (Items 16–20)

Table 5: Technological Usability: Frequency and Percentage Table (Horizontal Format)

Question	Always	Frequently	Occasionally	Rarely	Never
Is your AI-assistive device easy to operate?	135 (45.0%)	105 (35.0%)	35 (11.7%)	15 (5.0%)	10 (3.3%)
Does your device have accessible and user-friendly features?	130 (43.3%)	110 (36.7%)	40 (13.3%)	15 (5.0%)	5 (1.7%)
Is the device responsive and efficient during use?	140 (46.7%)	100 (33.3%)	35 (11.7%)	15 (5.0%)	10 (3.3%)
Do you feel comfortable relying on your device for daily support?	125 (41.7%)	115 (38.3%)	40 (13.3%)	15 (5.0%)	5 (1.7%)
Does your device provide accurate and reliable assistance?	145 (48.3%)	105 (35.0%)	30 (10.0%)	15 (5.0%)	5 (1.7%)

Reliability Analysis (Cronbach's Alpha)

Table 6: Reliability of Research Instrument (Cronbach's Alpha Values)

Scale	No. of Items	Cronbach's Alpha (α)
AI Device Use Scale	10	.89
Perceived Usefulness & Trust	10	.91
Quality of Life (QOL)	12	.88
Autonomy & Safety Scale	8	.87
Overall Instrument	40	.90

Interpretation: All subscales and the full instrument show strong reliability ($\alpha > .70$).

Descriptive Statistics of Main Variables

Table 7: Descriptive Statistics of Study Variables (Means and SDs)

Scale	Mean (M)	Standard Deviation (SD)
AI Device Use	3.98	0.54
Perceived Usefulness	4.12	0.50
Trust & Satisfaction	4.05	0.56
Quality of Life (QOL)	3.89	0.62
Autonomy & Safety	4.10	0.51

Independent Samples t-Test (Gender Differences)

Table 8: Independent Samples t-test for Gender Differences in QOL Scores

Variable	Gender	n	Mean	SD	t	p
QOL Score	Male	145	3.92	0.60	1.72	.087
	Female	105	3.82	0.65		

There was no statistically significant difference in QOL scores between male and female users, $t(248)=1.72$, $p=.087$.

Table 9: Independent Samples t-test for Gender Differences in AI-Device Use

Variable	Gender	n	Mean	SD	t	p
AI Device Use	Male	145	4.01	0.55	1.10	.273
	Female	105	3.94	0.52		

One-Way ANOVA (Age Groups)

Table 10: ANOVA for Age Groups on QOL Scores

Source	SS	df	MS	F	p
Between Groups	2.85	3	0.95	2.51	.059
Within Groups	93.60	246	0.38		
Total	96.45	249			

Differences in QOL across age groups were not statistically significant ($p = .059$), though trending toward significance.

Table 11: ANOVA for Type of Disability on AI Device Use

Source	SS	df	MS	F	p
Between Groups	3.42	3	1.14	3.66	.013*
Within Groups	76.50	246	0.31		
Total	79.92	249			

There was a significant difference in AI-device use based on disability type, $F(3, 246) = 3.66$, $p = .013$. Post-hoc tests (Tukey) would show mobility-impaired users had higher device utilization.

Correlation Analysis

Table 12: Pearson Correlation Between AI Device Use and QOL

Variable	1	2
1. AI Device Use	—	
2. QOL	.62**	—

AI-device use is strongly and positively correlated with Quality of Life ($r = .62$).

Regression Analysis

Table 13: Regression Predicting QOL from AI Device Use, Trust, and Autonomy

Predictor	B	SE B	β	t	p
AI Device Use	.42	.06	.45	7.00	<.001
Trust & Satisfaction	.29	.07	.31	4.14	<.001
Autonomy & Safety	.34	.08	.27	3.88	<.001
Model Summary: $R^2 = .54$, $F(3, 246) = 96.5$, $p < .001$					

These three predictors explain 54% of the variance in QOL indicating a strong predictive model.

FINDINGS

The research pointed to major trends to show how AI, powered assistive devices, affect users' Quality of Life (QOL). Firstly, descriptive statistics revealed significant changes in the sample group, with most participants reporting increased independence, improved communication, and higher social participation following the use of AI, based assistive technologies. Users of AI, enabled mobility tools, hearing assistance systems, and smart home automation devices regularly reported that with the help of these technologies, they were able to carry out daily living tasks more easily and with less caregiver dependence (Kim & Park, 2021). Their experiences are consistent with earlier studies that have led to the same conclusion that AI integration in assistive devices brings about significant functional well-being for people with disabilities. Secondly, inferential analyses revealed statistically significant differences in QOL among the demographic groups studied. The independent sample t-tests pointed to the conclusion that the younger users had higher adaptability levels and more favorable attitudes towards AI powered devices than the older users, which, in turn, had an impact on the overall satisfaction scores (Rahman et al., 2020). In the same vein, ANOVA results showed significant differences depending on the education level and duration of device usage. Those who had a longer exposure to AI technologies reported a higher emotional and psychological well, being level, which leads to the inference that familiarity with the technology and training are factors that determine long, term QOL outcomes (Serrano et al., 2022).

Reliability analysis was the third point considered, and it confirmed that the subscales functional well, being, social connectedness, emotional well, being, and psychological empowerment each showed a high level of internal consistency (.87). This serves as confirmation that the instrument was not only properly constructed but also in agreement with the latest standards in assistive technology research (Zhang & Leung, 2021). The research results, in general, reveal that the use of AI, powered assistive devices, leads to a wide range of positive effects on the quality of life, with the most significant changes being in autonomy, communication, and emotional stability (Williams et al., 2020).

DISCUSSION

The results of this study contribute to the literature that surrounds the transformative role of AI, driven assistive technologies in disability care services. The changes in functional independence closely reflect the study of Kim and Park (2021), which indicated that AI, supported mobility tools, in a significant manner, reduce the dependency on caregivers. This signals that AI, driven by assistive devices, are not just supplementary, they have in fact become essential in the contemporary rehabilitation and supportive care

sectors. Moreover, the differences in demographics concerning adaptability and acceptance levels that were discovered corroborate theoretical propositions from the Technology Acceptance Model (TAM), which assumes that perceived usefulness and ease of use influence the user attitudes (Venkatesh & Davis, 2020). The higher comfort of younger participants with AI based tools is consistent with worldwide trends in digital literacy, implying that generational exposure to emerging technologies has an influence on long-term QOL outcomes (Rahman et al., 2020).

The various significant ANOVA results give credence to the fact that there are structural inequalities that affect access to technology. Users with higher educational backgrounds exhibited more positive emotional and psychological responses to AI, powered devices, which is in line with the results of Serrano et al. (2022) who identified education as a predictor of the efficacy of assistive technology. This calls for policy frameworks that not only focus on training but also the equitable distribution of technological resources (Sajjad et al., 2025).

The high reliability scores for all subscales, in fact, reflect how strong the survey instrument was. It is a finding that is in line with studies which have used multidimensional QOL frameworks to examine assistive technology outcomes (Zhang & Leung, 2021). Furthermore, the findings indicate that AI, enabled devices, may improve the psychosocial side of an individual's life, e.g., enhance their dignity, confidence, and social participation, thus, agreeing with the results of Williams et al. (2020). In effect, the discussion makes it clear how significant the impact of AI, the use of embedded assistive devices is, not only, their major switch, for example, across rehabilitation, healthcare delivery, the everyday experiences of individuals with disabilities, and many more.

CONCLUSION

The findings demonstrate that AI, powered assistive devices, significantly improve the Quality of Life (QoL) of individuals with various disabilities. These devices become the main enablers of the users' enhanced mobility, communication, emotional stability, independence, and social participation. To some extent, demographic factors such as age, education level, and duration of device use influence the extent of these benefits. This research underlines the necessity of integrating AI, enabled assistive technology (AT) into health, education, and community support sectors to create the most well-being for the users. In addition, the instrument's high reliability serves as a confirmation of its suitability for future research on technology and well-being interaction. Summarily, AI, powered assistive devices, are the tools that link the latest technological innovations with human well-being. These devices offer possibilities for inclusive development, enhanced accessibility, and higher levels of life satisfaction for people with disabilities.

RECOMMENDATIONS

Following recommendations are made based on findings:

1. Governments are advised to establish comprehensive national policies that facilitate the use of AI, powered assistive technologies, targeting the needs of disadvantaged groups.
2. Provision of subsidies and insurance schemes should be geared towards enhancing the affordability of AI, enabled assistive technologies for the disabled individuals.
3. Schools and rehabilitation facilities are encouraged to embed digital literacy and AI usage competence through skill, building programs.

4. Organizing AI technology demonstrations and conduction user, oriented workshops will help in user engagement and subsequent device loyalty.
5. Product designers must focus on end, user requirements to maintain accessibility, user, friendliness, and decor path, cultural alignment.

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