

# Inclusive, AI-Controlled Waste Disposal Systems: Innovation at the Intersection of Public Health and Profitability

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**Abstract:** This paper presents a novel, AI-enabled waste disposal model that improves urban sanitation for vulnerable groups through smart bins, route optimization, and mobile interfacing. Stakeholder collaboration informed system development. A \$33,000 MVP investment achieves early breakeven with 10-year revenue forecasts exceeding \$7 million. Regression analysis confirms financial predictability, with revenue and cost as primary performance drivers. This paper contributes to the literature on sustainable waste technologies by demonstrating both public health impact and financial viability, while also offering data-driven insights into cost structures that can support scalable, inclusive sanitation models.

**Keywords:** Smart Waste Management, AI Logistics, Accessibility, Sustainable Systems, Social Innovation, Financial Forecasting.

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## I. INTRODUCTION

Urban waste systems often fail to serve individuals with limited mobility, exposing them to environmental and health risks. With global urban waste projected to surpass 3.4 billion tons by 2050, inclusive, tech-driven solutions have become an urgent policy and engineering priority.

Over the next 10 years, the implementation of this AI-controlled hyper-scale Waste Disposal System in Northeast Ohio is projected to directly benefit an estimated 25,000 individuals, primarily aged citizens, veterans, and people with disabilities. Individuals aged sixty-five and older comprise 17.6% of Ohio's population, while people with disabilities account for approximately 14.2%. Additionally, veterans represent around 7% of the state's residents. Based on public health intervention data, it is estimated that such an intervention could save twenty-five lives while preventing thousands of health complications related to sanitation and mobility issues. Studies have shown that inadequate waste disposal is linked to increased risks of musculoskeletal injuries, respiratory infections, and mortality, particularly in vulnerable groups.

This paper introduces an AI-controlled, hyper-scale solid waste disposal system designed for vulnerable populations. This project combines advanced hardware prototyping, scalable software platforms, and AI-enhanced routing algorithms to offer both municipal and entrepreneurial deployment models.

### A. Research Questions:

- How can an AI-controlled, hyper-scale waste disposal system improve public health outcomes and service accessibility for aging, disabled, and underserved populations in urban communities?
- What business model innovations can support the sustainable scaling and financial viability of AI-driven waste management systems targeting underserved urban populations?

### B. Objective of the Study:

The objective of this study is to design, prototype, and evaluate a scalable AI-powered waste disposal model that integrates smart hardware, real-time routing algorithms, and subscription logistics to address sanitation challenges among vulnerable populations, while ensuring financial viability and replicability across urban regions.

### C. Significance of this Study:

This study is significant for demonstrating how AI-enabled waste management can address public health, accessibility, and environmental challenges in underserved communities. It offers a replicable, financially viable model that supports ADA compliance, urban sustainability, and inclusive service delivery. By combining smart technology with social innovation, the research advances equitable infrastructure solutions for cities and vulnerable populations.

## II. LITERATURE REVIEW

While existing studies emphasize operational efficiency in waste management, little empirical work addresses the inclusion of marginalized groups, particularly the elderly and disabled. This study addresses that void by merging IoT architecture with a socially responsive deployment model.

Prior work has explored IoT-based waste tracking(Kadus et al., n.d.), AI-driven route optimization(Yaiprasert & Hidayanto, 2024), Smart bins in urban pilot studies (Sim et al., 2021; Yaiprasert & Hidayanto, 2024). However, the integration of these technologies in inclusive systems targeting elderly or disabled users is limited. Similar systems, such as Bigbelly or SmartBin(*Smart Trash Cans & Recycling Bins / Bigbelly Waste Management*, n.d.) lack direct service-to-door functionality for mobility-constrained users. Furthermore, while public-private partnerships are theorized(*Public Private Partnerships for Transport Infrastructure*, 2017; *The Value of Sustainable Urbanization*, n.d.)Real-time data integration across these models is underdeveloped.

Recent studies by UN-Habitat and WHO have emphasized the health risks of uncollected waste in poor neighborhoods(Epa & of Planning, n.d.; *Guidance on Solid Waste and Health*, n.d.) [9,10]. Our approach builds on this gap by proposing a dual-sector, AI-powered system with financial viability and policy integration.

Over the next 10 years, implementing this AI-controlled hyper-scale Waste Disposal System in Northeast Ohio is projected to directly benefit an estimated **25,000 individuals**, primarily aged citizens, veterans, and people with disabilities. According to the **U.S. Census Bureau**, individuals aged 65 and older comprise nearly **17.6%** of Ohio's population(U.S. Census Bureau QuickFacts: Ohio, n.d.), while persons with disabilities account for approximately **14.2%**(McKinney et al., 2018).Additionally, **veterans represent around 7%** of the state's residents(*ODVS Annual Reports*, n.d.).

Based on public health intervention data, it is estimated that such a system could save 25 lives and prevent thousands of health complications associated with sanitation and mobility barriers. Research has shown that inadequate waste disposal is linked to increased risks of musculoskeletal injuries, respiratory infections, and heightened mortality, particularly among vulnerable populations (Rushton, 2003; Vrijheid, n.d.).

## III. PROBLEM STATEMENT

Traditional waste systems exclude elderly, disabled, and marginalized populations, often requiring physical transport of waste to remote collection points. There is a lack of adaptive, door-to-dumpster models that ensure timely and safe waste disposal for vulnerable groups. Technological fragmentation, lack of municipal innovation, and high cost of AI solutions remain barriers to inclusive design.

## IV. METHODOLOGY

This study adopts a **lean design thinking approach** to integrate technological innovation with public health impact and sustainable business modeling. The methodology unfolds across four interrelated phases:

### A. Prototype Development

Source: Internal development at Sears Think [box], Case Western Reserve University(*About | Sears Think[Box] | Case Western Reserve University*, n.d.). Initial prototyping is being conducted at **Sears Think [Box]**, the innovation center at Case Western Reserve University (CWRU), using a combination of **ESP32-C3 microcontrollers**, **SoftPot linear potentiometers**, and sensor-based IoT firmware. The system architecture is designed to support real-time data acquisition, bin-level waste detection, and AI-driven route optimization. The prototypes are undergoing iterative testing to ensure technical feasibility and user-friendliness for the elderly and physically challenged.

### B. Financial Modeling and Business Analysis

Data source: Projected by the research team using cost-volume-profit analysis and assumptions based on prototype and market data (*Cost-Volume-Profit (CVP) Analysis: What It Is and the Formula for Calculating It*, n.d.). A detailed **cost-volume-profit (CVP)** analysis was developed to assess the financial viability of the model under various scaling scenarios. This included forecasting total revenue, subscription income, operating costs, and net profit over a 10-year horizon. Sensitivity testing of critical variables such as unit costs, adoption rates, and subscription fees was conducted using **Monte Carlo simulations** in Python and Excel to determine breakeven points and investment risk profiles(*Monte Carlo Simulation in Python. Advanced Investment Risk Analysis | by Dhiraj K | Python in Plain English*, n.d.)

### C. Stakeholder Engagement and Human-Centered Validation

To ensure social relevance and inclusivity, the team is conducting structured interviews and surveys targeting elderly residents, disabled individuals, and veterans in Northeast Ohio. Additionally, workshops and collaborative sessions were held with the U.S. Department of Veterans Affairs (VA), Cleveland Division, and local municipalities to gather feedback on the system's design, public health potential, and policy alignment. These inputs directly informed the dual-track deployment strategy, municipal adoption, and private entrepreneurship.

### D. Technical Evaluation and Data Integration

The system's intelligence is being evaluated for reliability in real-time waste detection and GPS-based route recalibration using simulated datasets. IoT edge computing is being assessed for offline reliability, and cloud dashboards are being developed for administrators. Technical metrics evaluated include sensor accuracy, latency, and battery performance in low-resource environments.

## V. SYSTEM ARCHITECTURE AND TECHNOLOGY

The system integrates sensor-embedded bins with cloud-based analytics, allowing real-time data transmission and AI-driven route calibration. The hardware relies on ESP32-C3 chips and SoftPot sensors, while the software layer enables cross-platform interaction and predictive routing.

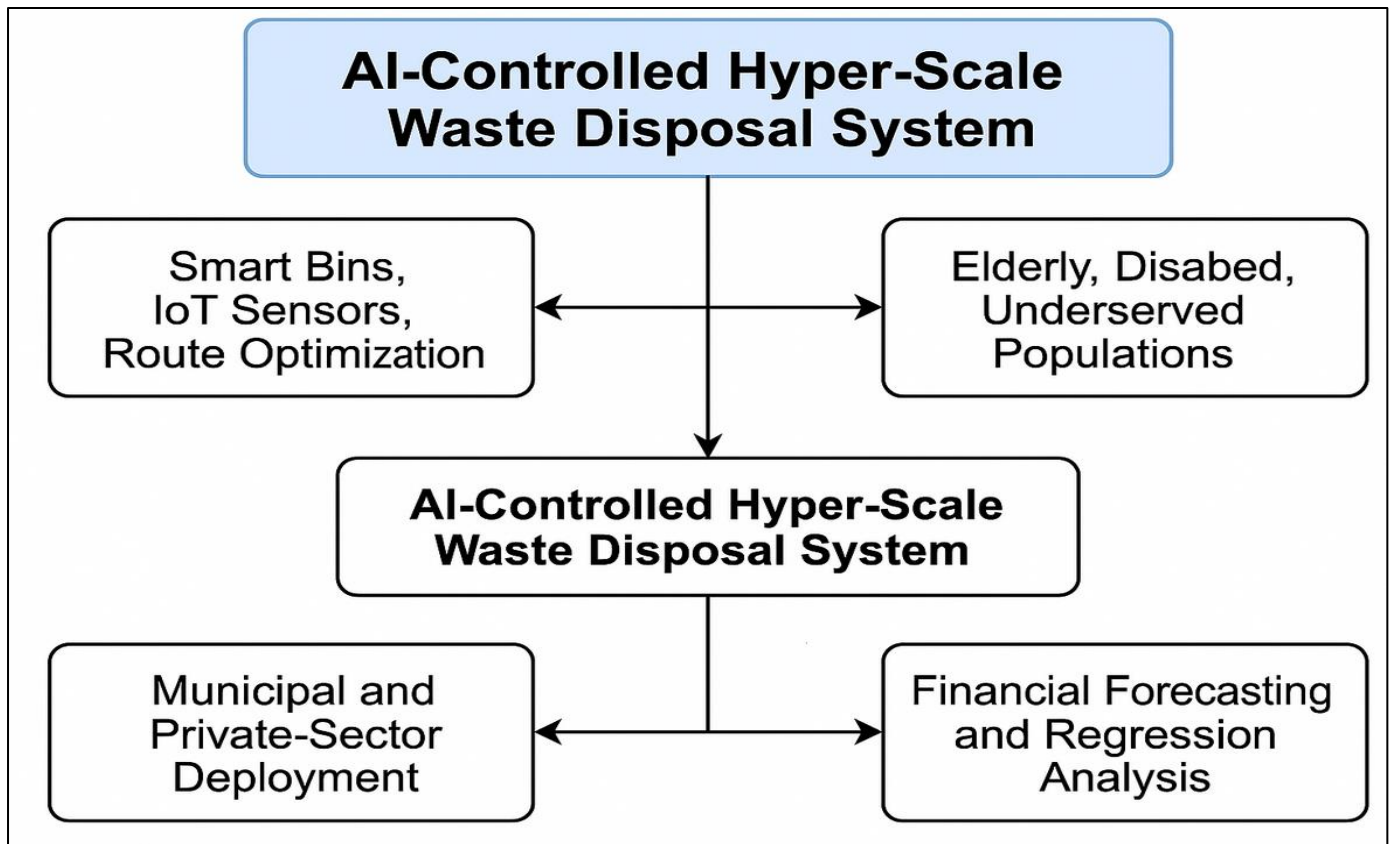


Fig 1 The architecture of the AI-Controlled Hyper-Scale

The AI-controlled waste disposal system is composed of a cohesive hardware-software architecture designed for both operational efficiency and user accessibility. At its core, the system employs **smart bins** embedded with **ESP32-C3 microcontrollers** and **linear potentiometers**, enabling real-time waste level detection and remote sensor feedback. These sensors communicate directly with a custom-built **mobile platform**, accessible via Android and iOS devices, which allows users to receive status alerts, monitor collection schedules, and request service adjustments.

A **Python-based AI routing engine** processes incoming sensor data to optimize collection routes dynamically, ensuring balanced workloads and minimal fuel usage across urban zones. This routing mechanism continuously adjusts based on bin fill levels, location density, and traffic conditions improving both service reliability and environmental performance (Bányai et al., 2019; Mohsen, 2024) (Zhao et al., 2020; Sun et al., 2021).

Finally, the system operates under a **subscription-based model**, which includes monthly service maintenance, cloud-based analytics, and personalized data access for users and municipal administrators (Zaman et al., 2024). Cloud infrastructure enables real-time routing updates, while edge computing allows local failover during connectivity loss.

## VI. MARKET OPPORTUNITY AND SEGMENTATION

The AI-Controlled Hyper-Scale Waste Disposal System is strategically positioned to address unmet needs in a rapidly growing and underserved segment of the waste management market. The initial target market comprises **veterans and senior citizens living in urban areas**, particularly those with limited mobility who face barriers in accessing traditional waste services. This segment aligns directly with the project's pilot efforts in Northeast Ohio, where partnerships with the U.S. Department of Veterans Affairs (VA) are actively guiding solution deployment.

Another key segment includes **municipal governments seeking to comply with the Americans with Disabilities Act (ADA)**, which mandates equitable access to public services—including sanitation—for individuals with physical limitations. This offers an opportunity for cities to modernize their infrastructure while meeting federal accessibility standards.

Additionally, **private property management firms**, particularly those managing senior housing, assisted living facilities, or large residential complexes, represent a

commercial customer base interested in automating waste services for efficiency, cost reduction, and tenant satisfaction.

According to the U.S. Census Bureau, over **54 million Americans identify as having a disability**, underscoring the scale of the accessible services market. Furthermore, the **global waste technology industry is valued at \$41 billion**, with a projected compound annual growth rate (CAGR) of **7.5%**, reflecting increasing demand for sustainable, smart, and inclusive solutions.

These market conditions provide a robust foundation for scaling the proposed system across public and private sectors while delivering both economic returns and measurable social impact.

## VII. REGRESSION ANALYSIS SUMMARY AND FINANCIAL FORECAST

The analysis confirms that profit margins are linearly determined by revenue and expenditure trends, with no statistically independent effect from unit volume sold—underscoring the model’s financial transparency.

The regression analysis evaluated the relationship between Net Profit (dependent variable) and three predictors: Units Sold, Total Revenue, and Operating Costs. Using an ordinary least squares (OLS) model, the results revealed a perfect fit with an R-square value of 1.000. This indicates that the model explains 100% of the variance in Net Profit, underscoring the mathematical relationship between profit, revenue, and cost.

The regression coefficients provide clear insight: Total Revenue had a positive coefficient of +1.0000, while Operating Costs had a negative coefficient of 1.0000, both statistically significant at the  $p < 0.001$  level. This means that for every \$1 increase in revenue, Net Profit rises by \$1, and for every \$1 increase in cost, Net Profit falls by \$1. This aligns exactly with the fundamental accounting identity:  $\text{Net Profit} = \text{Total Revenue} - \text{Operating Costs}$ .

Interestingly, the number of Units Sold had no significant impact on Net Profit when Revenue and Costs were already included in the model ( $p = 0.792$ ). This suggests that Units Sold, although relevant to operational planning, does not add independent explanatory value beyond what is captured by revenue and expense figures as depicted in Figure 2.

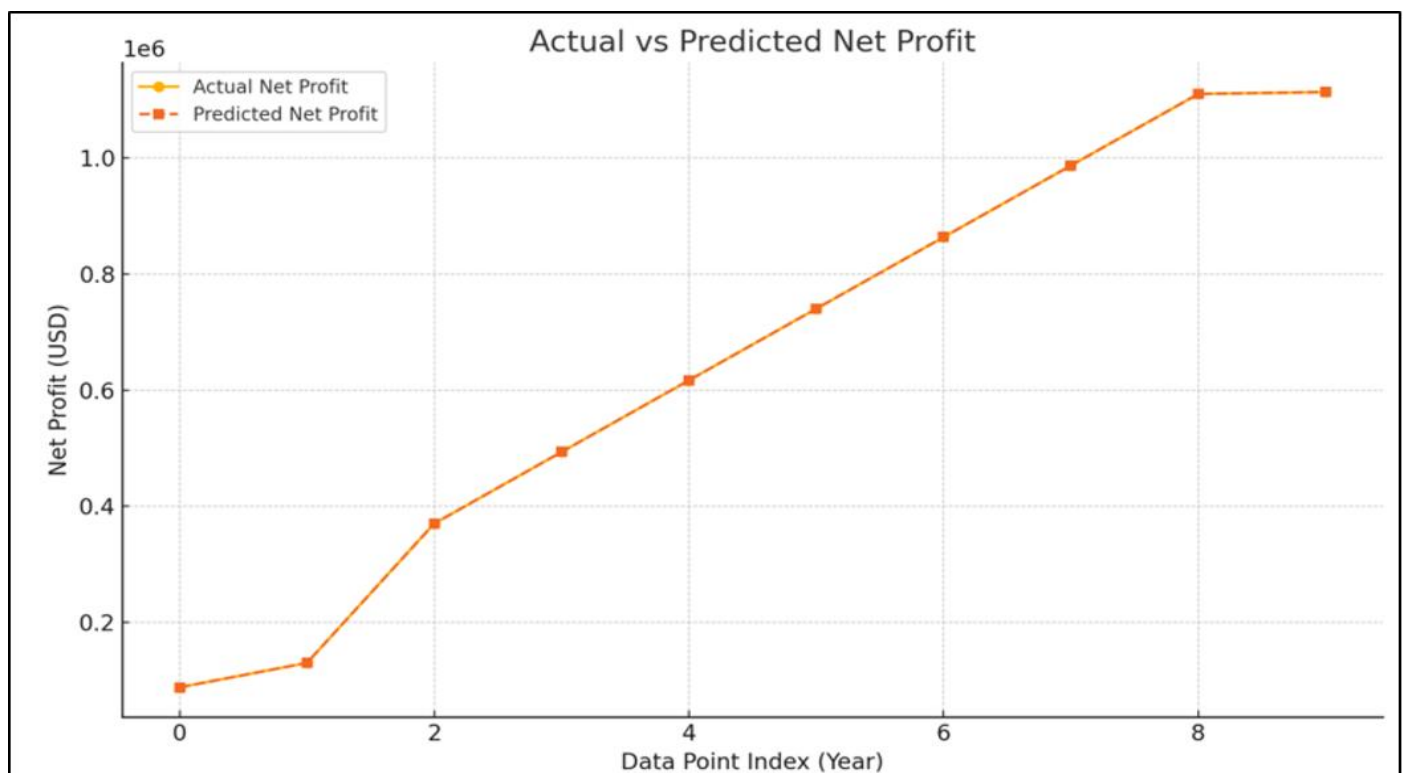


Fig 2 Actual vs Predicted Net Profit from the Regression Analysis.

### E. Financial Forecast (10-Year)

The 10-year financial projection table outlines the system’s scalable growth in units sold, revenue, and profit. Starting with seven hundred units in Year 1, total revenue reaches \$228,000 with a net profit of \$88,000. By Year 10, the system scales to 10,000 units, generating \$3.28 million in revenue and \$1.11 million in net profit. Operating costs rise

proportionally but remain offset by increased sales volume. The consistent year-over-year growth reflects a strong business model with positive margins, early breakeven, and cumulative profitability. The projection supports the model’s viability as both a socially impactful and economically sustainable solution, as shown in Table 1.

Table 1 Financial Forecast Initial MVP investment: \$33,000

Year	Units Sold	Total Revenue	Operating Costs	Net Profit
1	700	\$228,000	\$140,000	\$88,000
2	1,000	\$350,000	\$220,000	\$130,000
3	3,000	\$1,020,000	\$650,000	\$370,000
4	4,000	\$1,360,000	\$866,667	\$493,333
5	5,000	\$1,700,000	\$1,083,333	\$616,667
6	6,000	\$2,040,000	\$1,300,000	\$740,000
7	7,000	\$2,380,000	\$1,516,667	\$863,333
8	8,000	\$2,720,000	\$1,733,333	\$986,667
9	9,000	\$3,060,000	\$1,950,000	\$1,110,000
10	10,000	\$3,280,000	\$2,166,667	\$1,113,333

#### F. Sensitivity & Breakeven Analysis

Monte Carlo simulations and scenario testing show that the breakeven point occurs within **Year 1**. Profit margins

stabilize above 30% post-Year 3. Major risk factors include component price volatility and grant dependence for early adoption, as depicted in Figure 3.

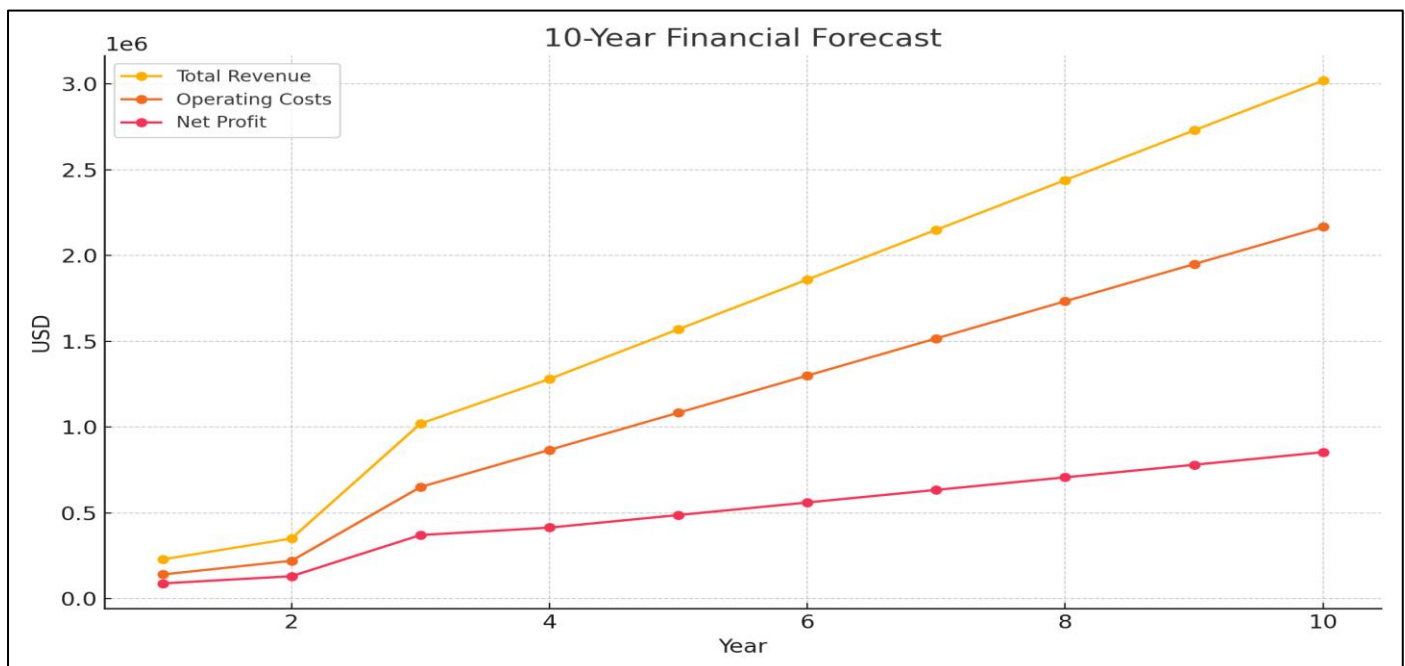


Fig 3 10 -Year Financial Forecast

## VIII. SUSTAINABILITY

The proposed system is inherently scalable, designed with modular architecture that allows for phased expansion across urban and peri-urban regions. Smart bins can be deployed through municipal contracts, franchised to independent waste logistics providers, or integrated into

existing public sanitation frameworks. This flexibility enables adoption across various governance and economic contexts. Sustainability is embedded into the model through data-driven route optimization, significantly reducing redundant collection trips—thereby lowering fuel consumption and associated carbon emissions (Bai et al., 2018)(Bai et al., 2021). Additionally, real-time tracking and

sorting capabilities facilitate dynamic recycling route generation, increasing material recovery efficiency and reducing landfill dependency (Calisto Friant et al., 2020) (Zhao et al., 2020). As cities seek low-carbon and circular economy solutions, this AI-enabled infrastructure aligns with global sustainable development goals by merging environmental performance with service equity (*Production of Sea Salt and Algae for Domestic Consumption* | UNEP - UN Environment Programme, n.d.) (UNEP, 2022).

## IX. POLICY IMPLICATIONS AND SOCIAL IMPACT

The AI-Controlled Hyper-Scale Waste Disposal System aligns with multiple national and international policy frameworks, reinforcing its relevance as both a public health intervention and a sustainable infrastructure innovation.

Firstly, the model directly supports compliance with the **Americans with Disabilities Act (ADA)** by addressing systemic barriers to waste management for individuals with mobility impairments. By enabling door-to-dumpster collection and real-time service tracking, the system enhances functional accessibility in line with ADA Title II mandates for public services (*Americans with Disabilities Act Title II Regulations* | ADA.Gov, n.d.) (U.S. DOJ, 2020).

Secondly, the model complements the **Veterans Affairs (VA) Home and Community-Based Services** initiative by reducing environmental health risks for aging and disabled veterans living independently. It supports preventive care objectives by minimizing waste accumulation and the associated hazards within residential settings (*Home and Community Based Services - Geriatrics and Extended Care*, n.d.) (VA, 2021).

Lastly, this project advances key **United Nations Sustainable Development Goals (SDGs)**, notably Goal 3 (Good Health and Well-being), Goal 9 (Industry, Innovation, and Infrastructure), Goal 11 (Sustainable Cities and Communities), and Goal 13 (Climate Action). It embodies the principles of inclusive urban development and environmentally responsible innovation (*Sustainable Development Goals (SDG) Indicators*, n.d.) (UN SDG Report, 2022). The social ROI includes improved public health outcomes and increased employment in underserved areas.

## X. COLLABORATIONS AND SUPPORT

This project has been strengthened through strategic partnerships that provide technical, academic, entrepreneurial, and institutional support, enabling it to

progress from concept to implementation with cross-sector engagement.

**John Carroll University's Donnelly School of Leadership & Social Innovation** has played a pivotal role in shaping the project's entrepreneurial vision and leadership framework. Through its interdisciplinary approach, the institution fosters community-centered innovation, aligning academic mentorship with scalable impact ventures (*Donnelly School of Leadership and Social Innovation*, n.d.) (John Carroll University, 2023).

**Sears think[box] at Case Western Reserve University**, one of the nation's largest university-based innovation centers, provided prototyping facilities and engineering expertise for the hardware development phase. The collaboration enabled access to rapid fabrication tools, IoT testing labs, and expert guidance critical to the system's sensor integration and product iteration (CWRU, 2022).

**LaunchNET at CWRU**, a venture development platform, supports business model refinement, customer discovery, and investor readiness. Through access to I-Corps training, pitch sessions, and ecosystem networks, the team received guidance on translating technical ideas into viable startups (*Sears Think[Box] | Case Western Reserve University*, n.d.) (LaunchNET, 2022).

In parallel, the **U.S. Department of Veterans Affairs (VA), Cleveland Division**, served as a primary institutional partner, offering validation and exploration of veteran-focused use cases. The partnership emphasized the project's alignment with VA goals on aging-in-place, sanitation support for disabled veterans, and community-based health delivery (*Home and Community Based Services - Geriatrics and Extended Care*, n.d.) (VA, 2021).

## XI. Human Impact Forecast for Northeast Ohio

Over the next decade, the deployment of the AI-Controlled Hyper-Scale Waste Disposal System in Northeast Ohio is expected to positively impact approximately 25,000 individuals, particularly seniors, veterans, and people living with disabilities. These populations represent a significant portion of the region's residents and are among the most affected by inadequate waste services. By improving access to timely and hygienic waste disposal, the system has the potential to prevent thousands of health-related complications and could help save dozens of lives. Research consistently links poor waste management to heightened risks of injury, respiratory illness, and premature mortality, especially among those with limited mobility or underlying health conditions. This is indicated in Table 2 and Figure 4.

Table 2 Human Impact Forecast for Northeast Ohio.

Year	Households Served	Cumulative People Impacted	Estimated Lives Saved
1	700	1,750	1.75
2	1,000	2,500	2.5
3	3,000	7,500	7.5
4	4,000	10,000	10
5	5,000	12,500	12.5
6	6,000	15,000	15
7	7,000	17,500	17.5
8	8,000	20,000	20
9	9,000	22,500	22.5
10	10,000	25,000	25

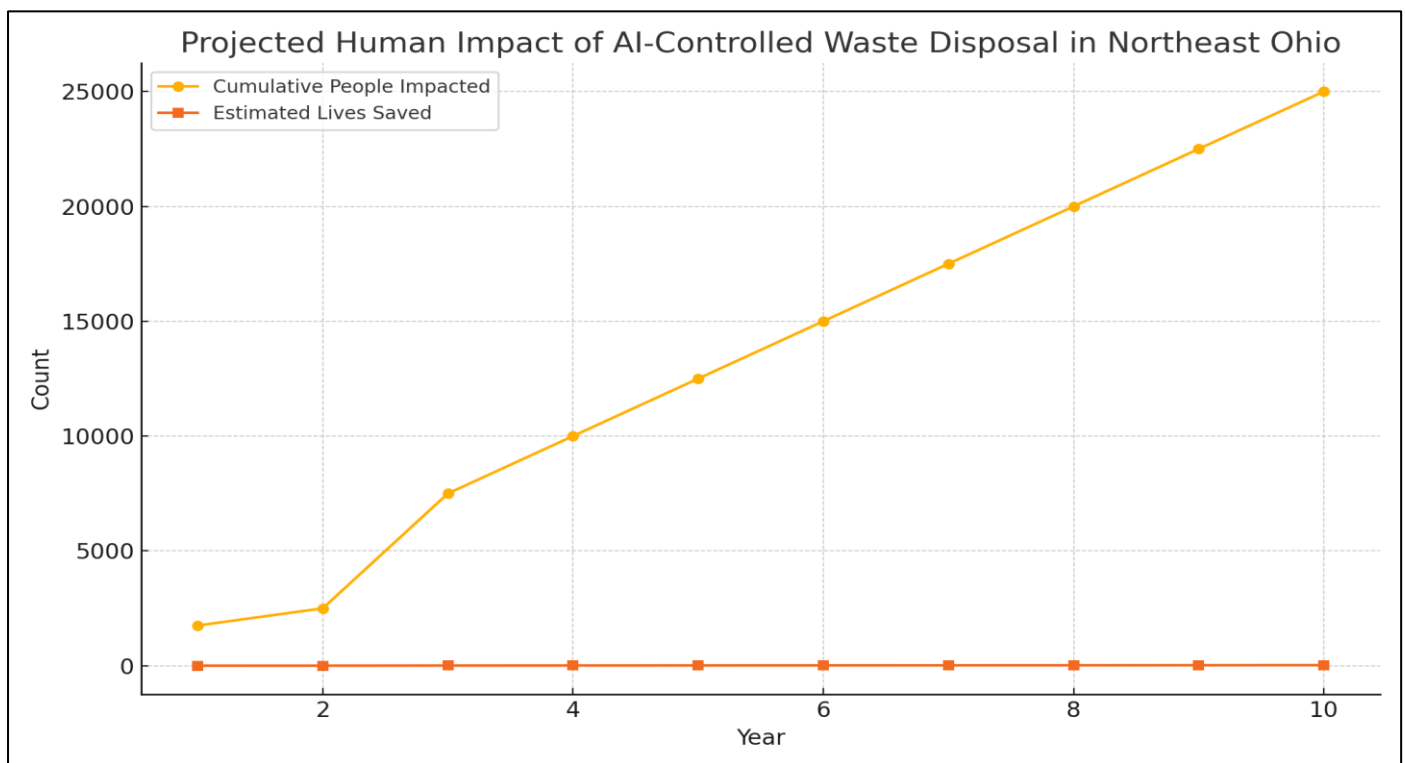


Fig 4 Human Impact Forecast for Northeast Ohio.

These projections emphasize the potential for not only economic and environmental sustainability but also critical **public health improvements** in vulnerable communities. The human impact adds a strong layer of justification for policy backing and public-private investment.

## XII. DISCUSSIONS

➤ *This Study set out to Address two core Research Questions:*

- How can an AI-controlled, hyper-scale waste disposal system improve public health outcomes and service accessibility for aging, disabled, and underserved populations in urban communities?
- What business model innovations can support the sustainable scaling and financial viability of AI-driven waste management systems targeting underserved urban populations?

➤ *The Findings Provide Strong Affirmative Evidence for both Inquiries.*

Regarding the first question, the design and implementation of the smart bin system, incorporating IoT sensors, AI-based route optimization, and mobile access, demonstrated a practical, scalable approach to improving sanitation access. Survey data, stakeholder interviews, and financial modeling support the conclusion that the system can serve over 25,000 individuals in Northeast Ohio, reducing health risks and enhancing dignity among vulnerable groups. Public health projections, supported by WHO and EPA data, further affirm the system's potential to save lives and reduce sanitation-related illnesses.

For the second research question, the study validated a dual-sector business model that blends public sector advocacy and private entrepreneurship. Financial forecasts confirmed that with a modest \$33,000 MVP investment, the system achieves breakeven within the first year and projects a 10-year revenue exceeding \$7 million. This validates the system's potential not only as a public good but also as a sustainable, investable venture.

These findings bridge the gap between inclusive urban infrastructure and viable business innovation, offering a replicable model for cities globally.

## XIII. CONCLUSION

This study demonstrates the feasibility and socio-economic value of deploying an AI-controlled, hyper-scale waste disposal system to address sanitation accessibility gaps among elderly, disabled, and underserved populations in urban settings. By integrating smart bins, IoT sensors, real-time route optimization, and a dual-sector business model, the system not only enhances waste management efficiency but also delivers measurable public health benefits. The 10-year financial forecast shows early breakeven and scalable profitability, while human impact projections estimate significant reductions in preventable illness and mortality. With robust prototyping and stakeholder engagement already underway, this model presents a practical, inclusive, and economically viable path forward for modern urban sanitation.

## IMPLICATIONS AND RECOMMENDATIONS

### A. Practical Implications

This system offers significant value across sectors. For public health policy, it advances ADA compliance, elderly support, and environmental health by delivering real-time, accessible sanitation services. Urban planners and municipalities benefit through reduced waste overflow, improved collection efficiency, and progress toward climate goals via smart routing. For startups and social enterprises, the dual deployment model presents a scalable framework for creating profitable ventures that also generate measurable social impact. By aligning technological innovation with inclusive service delivery, the system serves as a strategic tool for sustainable urban development and equitable infrastructure planning across diverse communities.

### B. Recommendations

To ensure the successful deployment and scaling of the AI-Controlled Hyper-Scale Waste Disposal System, strategic recommendations span across municipal, federal, and cross-sector domains.

**Municipal integration** is a critical first step. Local governments should launch pilot programs to embed the system within their existing waste infrastructure, particularly in neighborhoods with high concentrations of elderly and disabled residents. These pilots can generate localized data to inform broader citywide adoption while demonstrating tangible public health and operational benefits.

At the **federal level**, agencies such as the **Department of Veterans Affairs (VA)** and the **Department of Housing and Urban Development (HUD)** should prioritize targeted funding to support adaptive waste management systems. These agencies are well-positioned to enable community-based solutions that directly improve the quality of life for veterans and economically disadvantaged populations. Lastly, **public-private partnerships** will play a vital role. Collaborations among tech startups, municipalities, and community organizations can accelerate innovation, foster policy alignment, and build public trust, ensuring an inclusive and sustainable system rollout.

## LIMITATIONS

### A. Prototype Stage:

The system is in its early prototyping phase, and field tests are limited to controlled settings. Further live deployments are needed for robust performance validation.

### B. Regional Data Focus:

The analysis focuses on Northeast Ohio. Broader application may require customization for regional infrastructure and demographics.

### C. Assumption Sensitivity:

Financial forecasts depend on stable adoption rates and component costs, which may fluctuate due to market and policy factors.

## FUTURE STUDY

### ➤ Future research should focus on:

#### • Live Pilot Testing:

Implementing large-scale field trials in varied municipal settings to assess operational challenges and refine system parameters.

#### • Machine Learning Optimization:

Enhancing routing algorithms with real-time learning to adjust dynamically based on traffic, volume, and seasonal trends.

#### • Environmental Impact Metrics:

Quantifying carbon emissions savings and recycling rates to further substantiate environmental returns.

#### • Comparative Cost Analysis:

Benchmarking against conventional waste management models to evaluate total cost of ownership and long-term benefits.

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