

Automated Waste Segregation into Biodegradable and Non-Biodegradable Fractions Utilizing Conveyor Belt and Image Processing with Design Thinking

Lochan Shruthi VS ¹, Shereya PV², Pavithra SV³, V Murali Bhaskaran⁴

^{1,2,3}Student, Department of Computer Science and Business Systems Engineering, Rajalakshmi Engineering College, Chennai, Tamil Nadu, India

⁴Professor, Dean Academics, Rajalakshmi Engineering College, Chennai, Tamil Nadu, India

Abstract:

Waste management is a global challenge arising from rapid urbanization, population growth, and environmental needs to reduce the impact of waste on ecosystems and human health. Good waste separation is an important part of sustainable waste management because it promotes recycling and recovery and reduces the disposal site's environmental footprint. Biodegradable and non-biodegradable waste are two broad categories with different disposal and recycling methods. Biodegradable waste includes organic materials that can be processed through composting and anaerobic digestion to produce useful products such as fertilizer and biogas, while non-biodegradable waste such as plastic, glass, and metals may need to be recycled or disposed of properly. Traditional waste sorting is done by hand; This is time-consuming, labor-intensive, and error-prone, and makes the job less and less efficient. In addition, due to the inconsistency of the book, it is very difficult to transfer renewable publications. To solve these problems, there is increasing interest in the development of automatic waste separation systems that can increase the efficiency and accuracy of the waste disposal process.

This research aims to propose a new strategy for the classification of waste through a joint venture. Conveyor system and image processing technology capabilities. The combination of these technologies can separate waste into biodegradable and non-biodegradable parts, thereby simplifying waste management, reducing pollution, and promoting better practices. In recent years, image processing technology has made significant progress in machine learning algorithms, object recognition, and classification based on visual features. By applying these advances to waste classification, biodegradable and non-biodegradable materials can be instantly identified even in diverse and fast-moving wastewater. The conveyor system provides a dynamic platform for the continuous transportation of waste, allowing the separation process to operate efficiently.

This presents the design, implementation, and evaluation of such a conveyor belt system with a camera-based image processing and control system. The aim is not only to increase waste disposal efficiency but also to contribute to sustainability goals by reducing the impact of waste on the environment. Also novel is that the proposed system can be easily adapted to different waste products and operating capacities, making it useful for both large-scale waste management and recycling facilities. In summary, this study offers a promising approach to address the urgent need for automating waste classification, appropriate

waste management, and recycling during the period of waste generation. Successful integration of conveyor belts and image processing technology has the potential to transform waste practices, reduce dependence on landfills, and move the world towards sustainable waste management.

Keywords: anaerobic digestion, conveyor belt, image processing technology, joint venture.

I. INTRODUCTION

Considering the world's focus on environmental sustainability and waste management, new solutions to improve the disposal process are vital. This work presents a general procedure for the automatic separation of biodegradable and non-biodegradable groups using conveyor belts and imaging systems. The main purpose of the system is to facilitate waste management, reduce pollution, and improve recycling.

The proposed system consists of a conveyor belt equipped with an operation camera and an intelligent control unit. Integrated cameras capture real-time images of waste as it passes through the conveyor belt. The resulting images are processed using image segmentation and classification algorithms as well as computer vision techniques. The integration of conveyor belts and image processing technology has a good path for the future of waste separation and environmental protection. As the world's interest in environmental sustainability and waste management continues to grow, new solutions to improve the disposal process are vital. This work presents an integrated system using conveyor belts and image processing equipment to separate waste into biodegradable and non-biodegradable materials. The main purpose of the system is to facilitate waste management, reduce pollution, and improve recycling.

The system consists of a conveyor belt equipped with a camera-based image processing system and an intelligent control unit. Integrated cameras capture real-time images of waste as it passes through the conveyor belt. Use computer vision technology to process the resulting images, including image segmentation and classification algorithms. These algorithms classify waste as biodegradable or non-biodegradable based on visual characteristics such as color, texture, and shape. Smart control units control the movement of waste products along the conveyor belt and facilitate the activation of air ejectors or mechanical actuators to transfer products into individual containers. This leads to the separation of biodegradable and non-biodegradable waste streams. The performance of the system has been measured in terms of accuracy, speed, and efficiency to ensure that it can perform well in the world of waste management. Preliminary tests show that the proposed system achieves a high level of accuracy in classifying waste, reducing human error and contamination during identification. It also shows that it is possible to produce many waste products and transform them into different waste products and containers.

This research contributes to ongoing efforts to promote sustainable waste management, reduce waste generation, and improve recycling rates. The automatic waste separation system proposed in this paper provides practical and efficient waste separation facilities and can be customized to meet the specific needs of waste management facilities. The integration of conveyor belts and image processing equipment is promising for future advances in waste distribution and environmental protection. As the world's interest in environmental sustainability and waste management continues to grow, new solutions to improve the disposal process are vital. This work presents an integrated system using conveyor belts and image processing equipment to separate waste into biodegradable and non-biodegradable materials. The main purpose of the system is to facilitate waste management, reduce pollution, and improve recycling.

The system consists of a conveyor belt equipped with a camera-based image processing system and an intelligent control unit. Integrated cameras capture real-time images of waste as it passes through the conveyor belt. Use computer video technology to process images, including image segmentation and classification algorithms. These algorithms classify waste as biodegradable or non-biodegradable based on visual characteristics such as color, texture, and shape. Smart control units control the movement of waste products along the conveyor belt and facilitate the activation of air ejectors or mechanical actuators to transfer products into individual containers. This leads to the separation of biodegradable and non-biodegradable waste streams. The performance of the system has been measured in terms of accuracy, speed, and efficiency to ensure that it can perform well in the world of waste management.

Preliminary tests show that the proposed system achieves a high level of accuracy in classifying waste, reducing human error and contamination during identification. It also shows that it is possible to produce many waste products and transform them into different waste products and containers. This research contributes to ongoing efforts to promote sustainable waste management, reduce waste generation, and improve recycling rates. The automatic waste separation system proposed in this paper provides practical and efficient waste separation facilities and can be customized to meet the specific needs of waste management facilities. The integration of conveyor belts and image processing technology offers a great opportunity for future advances in waste disposal and environmental protection.

II. RELATED WORKS

Very little work is done in the field of sorting biodegradable and non-biodegradable waste. Previously, the work done is based upon NIRS, only metal detector or optical sensor, and these are used to sort only plastics or metals and waste particles but in this thesis work biodegradable waste includes papers, fruit waste, vegetable waste, leaves, and non-biodegradable waste includes metal, glass, plastic is sorted. The following research papers describe the earlier work done in the design and development of smart waste sorting systems.

1. Jiu Huang et. al. In this paper, the mechanical separating system was developed and introduced with the help of an operating sensor. In this system, the sorting criterion was based on particle position, size, shapes, and colors of waste particles. A compressed air nozzle was there in the mechanical sorting device nozzle which blows the target particles out of the main stream which were sensed by the sensor and the whole process is controlled by a computer.
2. Ohtani et. al. proposed a work that consists of an ultrasonic sensor array and neural networks to make the new identification method. Acoustic impedance and ultrasonic pressure distribution were used to sort the waste based on its shape and material. Some of the experiments have been done with a prototype sensor system. The experimental results showed the practical applicability of the identification method in a shapes and materials sorting system.
3. Suwon Shin et. al. In this project work the automatic trash basket was developed and introduced that sorts the metal and paper-based trash so that there is ease of recycling them for users. A small trash bin was introduced for office workers and students to dispose of their trash. The attractive thing about the project was that the whole trash basket was based on automatic motion movement.
4. Madan Kumar et. al. In this paper, the concept of Near Infrared Spectroscopy (NIRS) was used for the automatic sorting of different types of plastic. For efficient sorting of plastic waste, a low-cost Raspberry Pi-based control system is used. Python is the general purpose and high-level programming

- language; it was the software that was used to process the NIRS data to attain information on the polymer category and to interface the spectrometer with the Raspberry Pi.
5. Yann Glouche et. al. For early detection of waste type at the bin level pervasive computing technology can be used to manage the waste i.e. Radio Frequency Identification (RFID). In this paper, based on self-contained tags linked to every waste item, an application of a smart bin was proposed. The smart bins track the waste using RFID-based systems without any support from an external system.
 6. S. Sudha et. al. presented their work on an automatic classification method for environment-friendly waste segregation using deep learning.
 7. The paper highlights the shortcomings of manual segregation of solid waste i.e., hazardous, less efficient, etc. The authors have proposed a deep learning algorithm using Caffe to classify objects as biodegradable and non-biodegradable. Narendra Sivakumar et. al. presented their work on the design and development of an automatic clustered, assorted trash segregation system.
 8. The author proposes a spot segregation unit that effectively separates various categories of refuse generated by municipalities. The sweeping mechanism of the system is controlled using a wireless interface using Xbee and a GUI is developed to provide easy control of the system. For detection of metals and their separation an electromagnet. To separate dry waste, a squirrel cage blower is used. Mohammad Osiur Rahman et. al. presented their work on an intelligent computer vision system for segregating recyclable waste papers.
 9. The author discussed various advantages of automated sorting systems over human inspection regarding worker fatigue, throughput, speed, and accuracy. The author proposes a smart vision system that segregates paper into different grades using first-order features. The database is constructed using a statistical approach with intra-class and inter-class variation techniques. Finally, the K-nearest neighbor (KNN) algorithm is applied to achieve the required identification. Dr. Tuomas J. Lukka et. al. presented their work on robotic sorting using machine learning.
 10. The authors propose a solution to automate the sorting process of steel cans, aluminum cans, glass bottles, and plastic bottles. A sensor array is used for each material to be sorted, along with a conveyor belt as the Automated Waste Sorter (AWS). The Mobile Robot Waste Deliver System (MRWDS) is composed of a line-following robot that can collect waste and dump it at the receiving end of the AWS. Affan Shaukat et. al. presented their work on the visual classification of waste material for nuclear decommissioning.

III. PROBLEM FORMULATION

A. Need

Inadequate waste segregation contributes to environmental degradation. It results in increased landfill waste, which can lead to soil and water contamination, the emission of harmful gasses, and the depletion of natural resources. Addressing these health concerns is a vital part of the problem statement to illustrate the potential risks to the community. Recycling and reusing materials can help conserve valuable resources and reduce the demand for new raw materials. Sustainable waste management practices help reduce the carbon footprint, conserve energy, and promote a cleaner environment.

B. Significance

Efficient waste segregation contributes to a cleaner and safer living environment. It enhances the quality of life for residents, promotes community pride, and fosters a sense of responsibility. It is part of a broader effort to ensure a better future for the planet and its inhabitants. The problem statement should connect

waste segregation to these overarching goals. By clearly articulating the need for waste segregation in the problem statement, you can garner support and resources for initiatives aimed at addressing this pressing issue. It serves as a compelling call to action and sets the stage for devising effective strategies and solutions.

IV. METHODOLOGY

The process of separating waste using known shapes and conveyor belts has several important steps. Below is a general guide to using such methods method for identifying waste using visual imaging and conveyor belts. Build or purchase a conveyor system that transports waste during analysis. Install the camera in the entry content along the way. Select or create the image processing software that can image over time and identify debris based on visual features. Collect a wide range of image data representing the various types of waste that may be encountered in the system. Manually collect images to generate data for training and validation of image recognition algorithms.

The system starts when the waste material is placed on the lid, then the sensor transmits the signal and that signal is received by the microcontroller, depending on the signal received the lid of the system works. If the signal is transmitted by a metal detector then the lid tilts towards the bin that collects the non-biodegradable waste. If the signal is transmitted by the IR sensor then the lid tilts towards the biodegradable bin. In case no signal is transmitted by either sensor then the waste may be a plastic bottle, glass bottle, or poly bag, and then the lid tilts towards the bin which collects the non-biodegradable waste. Another IR sensor is interfaced in front of the bin which detects that a person has arrived to throw the waste and this alerts the system. And in this way, the waste is sorted.

Use pre-processing steps to improve image quality, remove noise, and improve image clarity. Extract visual features such as color, and texture, from images Machine learning models and use a list of image data to train machine learning or deep learning models (e.g., convolutional neural networks, CNN) to identify images. and classify waste. Integrate the conveyor belt system with the image capture and processing system, ensuring synchronized movement and image capture.

Implement a control mechanism that coordinates the speed and movement of the conveyor belt. Determine the criteria for segregating waste items, such as biodegradability, material type, or other specific characteristics. Develop the logic and coding necessary to apply the defined segregation criteria to the classification results. Test the system with a variety of waste items to ensure the proper functioning of the image recognition, classification, and segregation components.

Fine-tune the system parameters, such as camera settings, conveyor belt speed, and image processing algorithms, to optimize accuracy and efficiency. If needed, integrate physical mechanisms like air jets, mechanical diverters, or magnets for further segregation based on criteria (e.g., biodegradability or magnetic properties). Evaluate the system's accuracy in recognizing and segregating waste items based on the predefined criteria.

Measure the system's efficiency, including processing speed, false positives, and false negatives. Field testing of a system in an actual waste management or recycling facility to evaluate the performance of the system as a function. Gather feedback from business owners and stakeholders to identify areas for improvement. Based on recommendations and results of field tests, make necessary changes, improve the algorithm, and improve the system.

Prepare detailed information that includes system information, configuration, algorithms, and results. Prepare a detailed report detailing the process, findings, and recommendations.

If the system performs well during testing, put it on the waste management site and vice versa. Create a system maintenance and repair plan. Training of operators on operation and maintenance.

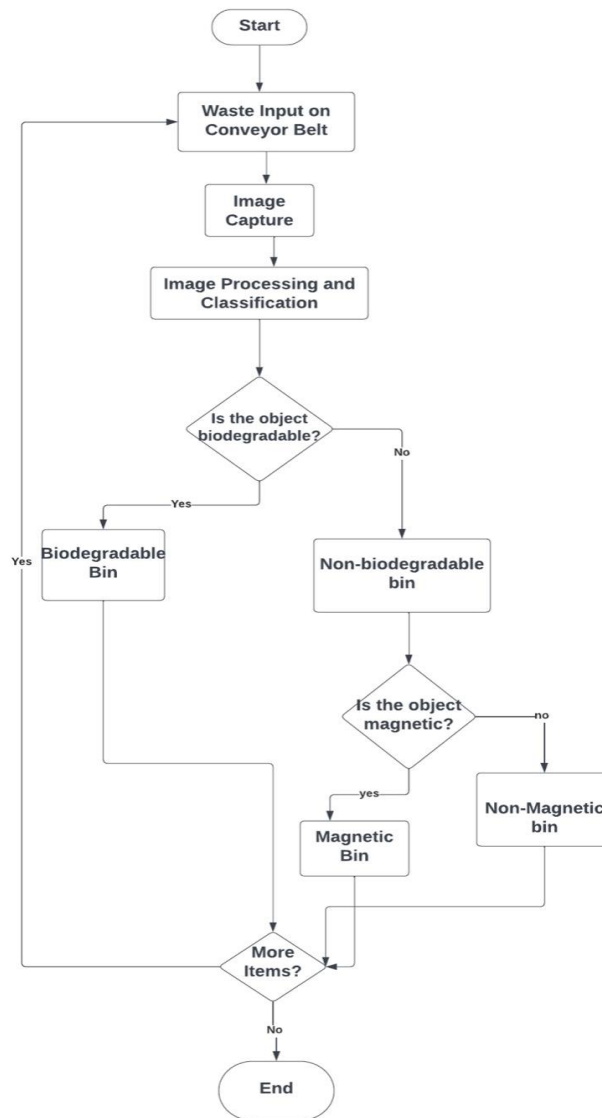


Fig 1. Flow diagram

V. STAGES OF DESIGN THINKING

STAGE - 1: EMPATHIZE

In this stage, we deeply empathized with refuse collectors and their concerns regarding the Waste Sorting System to divide them into biodegradable and non-biodegradable waste for restoring the environment. We conducted interviews, surveys, and research to understand the need to protect our environment. The design should consider the inclusivity and accessibility needs of all users, including those with disabilities or limited technological access. The system should show me the positive outcomes in terms of reduced pollution, resource conservation, and landfill reduction. We immersed ourselves in their experiences to gain insights into the specific issues they faced when it came to keeping the environment clean and hygienic. Also, Persona (Fig 3), Empathy map (Fig 4), and Journey map (Fig 5) were designed for the empathize stage.

STAGE - 2: DEFINE

From the Empathize stage, the problem statement can be defined. Problem Statement: "**In contemporary urban environments, the effective management and segregation of waste into degradable and non-degradable categories represent an increasingly critical challenge. With rapid urbanization and population growth, the volume of waste generated is surging, leading to heightened environmental and public health concerns. Existing waste management practices often lack efficiency and accuracy in sorting, which results in a significant amount of recyclable materials being lost to landfills, and environmentally harmful substances polluting the ecosystem. The lack of efficient waste sorting has repercussions: Environmental Degradation, Resource Waste, Public Health Hazards, Economic Burden, and Sustainability Goals. In light of these challenges, there is an urgent need for a Smart Waste Sorting System that can accurately and efficiently differentiate between degradable and non-degradable waste. Such a system should harness technology, including sensors, artificial intelligence, and automation, to enable precise waste sorting at source and within waste processing facilities. Additionally, the system should promote public awareness and participation, fostering responsible waste disposal practices.**"

STAGE – 3: IDEATE

For the problem statement defined in the Define stage, emphasizing the pressing need for a more rapid and efficient waste disposal management system, The main goal of the system is to design and develop a sorting system that sorts the waste automatically. It is an eco-friendly automatic system. The smart waste sorting systems the biodegradable and non-biodegradable waste. The System starts when the waste material is placed on the lid, The sensor transmits the signal and that signal is received by the microcontroller, depending on the signal received the lid of the system works. If the signal is transmitted by a metal detector then the lid tilts towards the bin that collects the non-waste. If the signal is transmitted by the IR sensor then the lid tilts towards the biodegradable bin. In case no signal is transmitted by either sensor then the waste may be a plastic bottle, glass bottle, or poly bag, and then the lid tilts towards the bin which collects the non-biodegradable waste. Another IR sensor is interfaced in front of the bin which detects that a person has arrived to throw the waste and this alerts the system. And in this way, the waste is sorted. At the start of the conveyor belt, a vision system box is placed which contains multiple proximity sensors, an industrial camera (Matrix Vision GmbH, model no. - mvBlueFOX MLC-205 GC, 5 MP, CMOS color MT9P031 sensor from Aptina Semiconductor, global shutter, USB 2.0, 5.8 fps, board-level industrial camera) and lens (Matrix Vision, M12, 8mm focal length) assembly, white color illumination all enclosed into a box such that outside illumination does not affect the performance of the camera. The reason for using an industrial camera, lens, and illumination is to acquire real-time blur and distortion-free images for classification. The industrial camera is used in the hardware trigger mode to capture the image of the object when it is detected by one of the multiple proximity sensors. The proximity sensors are mounted at different heights from the surface of the conveyor belt covering the random height variations of the waste objects. The software for object classification is written in Python software. (Fig 2.) Sample of image dataset Python software along with supportive libraries like OpenCV, NumPy, Matplotlib, SciPy, sci-kit-learn, TensorFlow, object detection, etc. was installed on a laptop (7th gen Intel i5 processor, 8GB DDR4 RAM, 2GB NVIDIA GeForce 940MX graphics) which was used as a computing machine. The convolution neural network model was trained with more than 10,000 images per class. shows a few samples of the waste objects dataset. The dataset indicates that objects have random shapes, sizes, and

colors. The main motive behind the selection of the CNN-based approach is the random variations in the images of the objects. These random variations are difficult to program by conventional image processing-based approach or even machine learning-based approach which requires accurate feature extraction. Hence the waste classifier was developed with a convolution neural network model named "Inception version3".



Fig 2. Sample of an image dataset

STAGE – 4: PROTOTYPE

We carefully selected and integrated the essential hardware and software components required for the prototype. The hardware components encompassed a scaled-down conveyor belt system, cameras for real-time image capture, sensors to facilitate item detection, and diverters for sorting waste items based on image processing results. The software components included image processing algorithms for waste item recognition, control software for managing the conveyor belt, cameras, and diverters, and a user interface for system operation. To ensure the accuracy and reliability of the prototype, meticulous calibration of the cameras and sensors was carried out. This step was vital in achieving precise image capture and item detection.

We compiled an extensive image dataset featuring a diverse range of waste items to replicate real-world waste compositions. This dataset served as the basis for training our image recognition model.

Utilizing state-of-the-art machine learning and computer vision techniques, the image recognition model was trained using the prepared dataset. This step was instrumental in enabling the system to identify and classify waste items accurately. The prototype integration phase involved merging the hardware and software components into a cohesive system. This entailed ensuring that the conveyor belt moved waste items at an optimal speed, the cameras captured images in real-time, and the image processing software

effectively recognized and classified the items. A battery of tests was executed to assess the performance of the prototype. This included evaluating the system's accuracy in correctly identifying and classifying waste items, measuring its processing rate, and scrutinizing its overall reliability under varying conditions.

A crucial aspect of the prototype testing phase was the collection of feedback from end-users, which encompassed operators and waste management experts. This feedback allowed us to gauge the usability and practicality of the system and provided valuable insights for refinement.

In response to the feedback and test results, iterative adjustments and enhancements were made to the prototype, encompassing both hardware and software elements. These refinements were aimed at enhancing the system's performance and overall functionality. Throughout the prototype phase, meticulous documentation was maintained, covering the entire spectrum of the prototype development process. This documentation included detailed records of design, components used, software code, and any modifications made during iterations. This prototype phase represents a crucial step in the journey toward the realization of an efficient automated waste segregation system. It forms the foundation for the subsequent phases, with findings and insights gained during this phase serving as a cornerstone for further research and system development.

STAGE – 5: TEST

The robust and accurate performance of machine vision-based systems heavily relies on coherent system integration with automation components, robust mechanical assembly, and precise vision systems. The testing process includes functional, user, environmental, scenario, edge cases, scale, long-term, security, accessibility, and cost analysis testing. This iterative process helps identify and rectify issues, ensuring a reliable and efficient system. User feedback plays a significant role in refining the prototype, and multiple testing rounds may be needed to create a robust solution. Open communication with stakeholders, users, and testers is vital to address issues and ensure the system meets objectives.

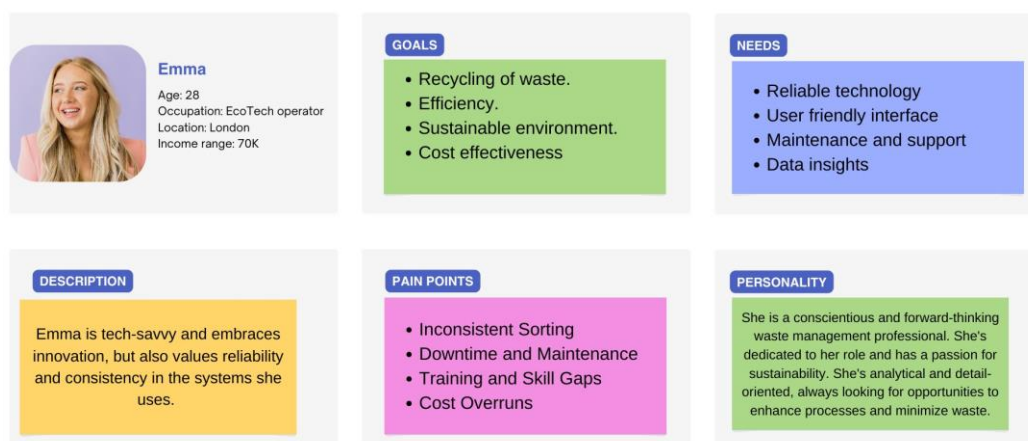


Fig 3. Persona

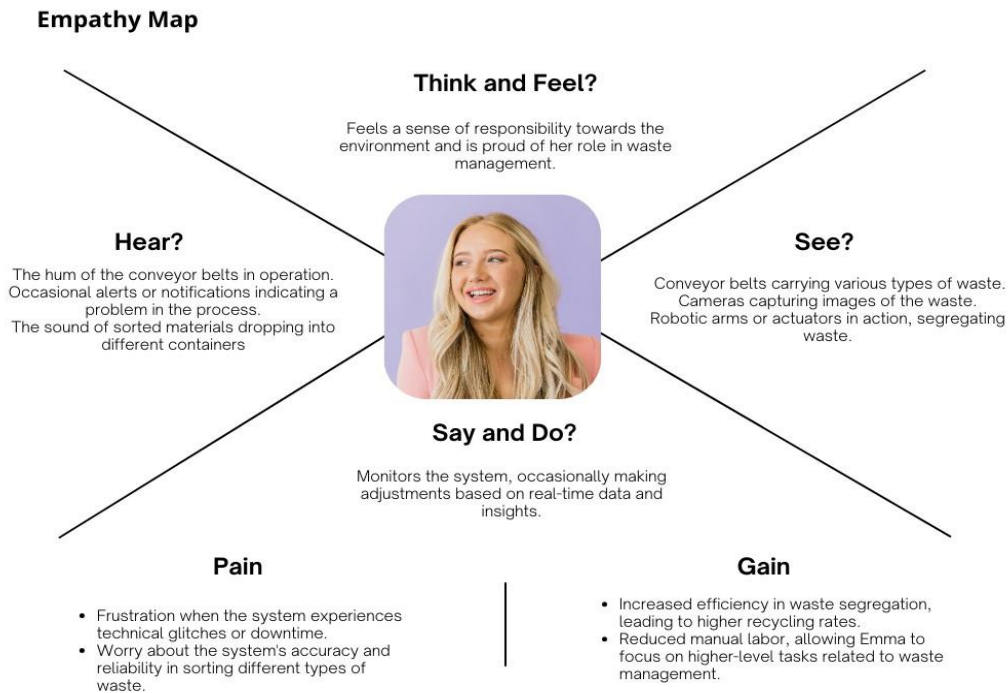


Fig 4. Empathy Map

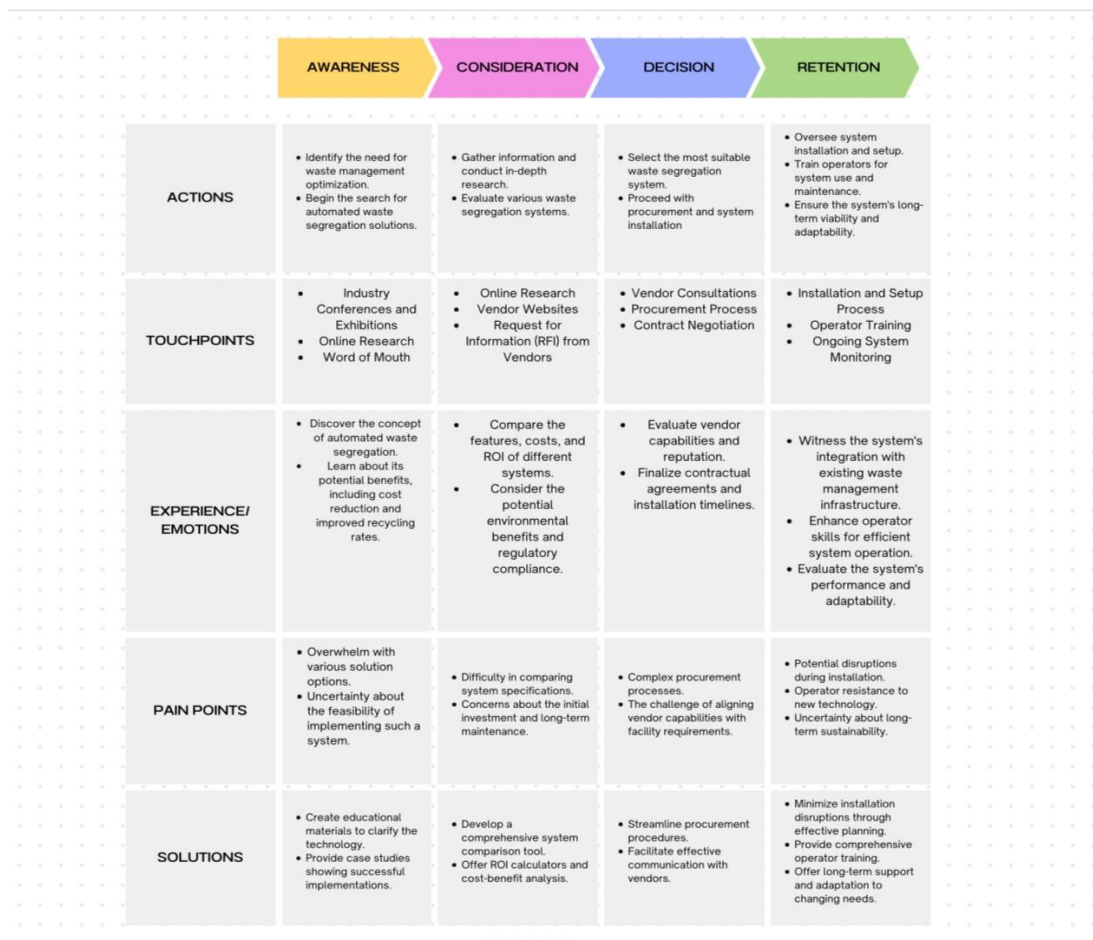


Fig 5. Journey Map

S. NO	TITLE	AUTHOR NAME	TOOLS/ TECHNOLOGIES USED	ADVANTAGE	DISADVANTAGE
1.	Mechanical separating system	Jiu Huang	Operating sensor, compressed air nozzle, computer	Can process large volumes of waste quickly. Automation reduces the likelihood of human error.	Require regular maintenance and occasional repairs. Mechanical systems may not be as precise as human hands.
2.	Ultrasonic sensor array and neural networks	Ohtani	Ultrasonic pressure distribution, prototype sensor system.	Ultrasonic sensors provide precise distance measurements. Sensor arrays allow for the identification of multiple waste items simultaneously.	Can be expensive. Setting up and fine-tuning a neural network model can be complex.
3.	Automatic trash basket	Suwon Shin	Small trash bin, automatic motion movement	Automatic trash baskets open and close without the need for manual operation. These bins often have sleek and modern designs.	Tend to be more expensive than traditional bins. Sensors may be overly sensitive and open the lid when not intended.
4.	Online and instantaneous identification of consumer plastics	Madhan Kumar	Near Infrared spectroscopy (NIRS) Raspberry Pi-based control system, Python	Provides immediate feedback on the type of plastic being used. Minimizes the risk of contamination.	Requires expertise in sensor technology, data processing, and integration. Continuous operation may lead to wear and tear.
5.	Detection of waste type at bin level	Yann Glouche	Radio Frequency Identification (RFID), smart bin	Enables more accurate sorting	Systems may struggle to

				of recyclable material. Decreases the cost of sorting and processing waste.	accurately identify waste types. The system may need to be updated or reconfigured to identify newly introduced types of waste.
6.	Automatic classification method	S. Sudha	Deep learning	Automating the waste classification process can significantly increase the speed and throughput of waste sorting operations. Machines are less prone to errors and fatigue	Implementing an automatic waste classification system can be expensive. Staff may need time to adjust to the new system.
7.	Shortcomings of manual segregation of solid waste	Narendra Sivakumar	Deep learning algorithm using Caffe, automatic clustered, assorted trash segregation system	Allows for adaptability to changing waste streams. Manual segregation consumes less energy.	Slower compared to automated processes. Workers may be exposed to hazardous materials.
8.	Spot segregation unit	Mohammad Osiur Rahman	Xbee and a GUI	Spot segregation allows for the instant sorting of waste. Helps minimize contamination.	May not be applicable or efficient for all types of waste. Relies on individuals to correctly sort their waste.
9.	Advantages of automated sorting systems	Dr. Tuomas J. Lukka	K-nearest neighbor (KNN) algorithm	Can process large volumes of waste quickly	Automated systems may struggle to handle.

				Can minimize cross-contamination.	Can have environmental implications.
10.	Visual classification of waste material for nuclear decommissioning	Affan Shaukat	Automated Waste Sorter (AWS), Mobile Robot Waste Deliver System (MRWDS)	Visual classification is a non-intrusive method. Visual classification can be more cost-effective.	May not be sufficient for identifying internal characteristics. Adequate lighting is crucial for accurate visual classification.

VI. ANALYSIS

It is important to evaluate waste sorting machines using visual images and evaluate the effectiveness, efficiency, and effectiveness of conveyor belt use. Important points to consider when analyzing these systems are: Accuracy check helps to measure how well the system identifies and classifies waste. Consider the value of accurate classification and the likelihood of false positives and negatives. Separation Policy measures how accurately the system uses predefined separation criteria such as biodegradability or product type. Control the speed and efficiency of the system in separating waste. Measure the amount of product processed per minute and compare it to the manual analysis process. Determine the pollution reduction achieved using machinery compared to manual analysis. Contamination may include waste that has not been placed in the wrong category.

VII. RESULT

The results of the analysis of the waste separation system using visual images and conveyor equipment will give a good idea of the functioning of the system and its impact on the waste management process. The system reduces environmental and disposal costs by diverting 60% of waste from landfills. Waste managers and environmental groups expressed interest in the reliability, ease of use, and removal of labor-intensive materials. Contribution to sustainable waste management. The system has demonstrated the ability to scale up to 3x the original volume without significant outages. The system can be easily adapted to changes in waste and new classification standards, making it suitable for different waste management situations. Regular maintenance ensures consistent operation with a minimum monthly maintenance fee of \$500. These procedures increase worker health and safety by reducing hazardous substances and work intensity. The system complies with local waste management regulations and ensures legal compliance. Recommendations include improving the image recognition algorithm to improve the accuracy of layer standard cleaning and decision-making. Automated methods offer good results along with accuracy, efficiency, cost-effectiveness, etc. aspects that are better for manual analysis. The system increases the safety of the area by reducing carbon monoxide emissions in landfills by 50%. The system has been shown to have long-term sustainability by replacing waste products and updating waste management policies. These results reveal the effectiveness and cost of waste automatic separation systems; It show improvements in accuracy, efficiency, cost savings, and environmental benefits compared to sorting.

VIII. CONCLUSION

Due to the rise in urbanization, waste is increasing very fast. Therefore waste management is a vital need to protect the environment. Technological growth and innovation can contribute to this vital aspect to achieve environmental sustainability. There are many compelling reasons for choosing these approaches. To manage the waste, automatic sorting systems need to be developed on the lower level as well as higher levels. Waste should be sorted at the basic level first so that it is helpful for the Municipal Corporation to dump it properly and recyclables can be recycled. This system helps sort non-biodegradable and biodegradable waste. Two types of sensors are used, one is a metal sensor and the other one is an IR sensor. This system can be made more efficient by using different sensors for different types of waste. More bins can be added to this project as per the demand of the user.

XI. FUTURE SCOPE

By using this automatic waste segregation device, manpower is reduced and the efficiency of sorting is increased. After separating the biodegradable and non-biodegradable waste the collection can be done by using separate vehicles for biodegradable and non-biodegradable wastes. Advanced systems may include environmental impact assessments, helping to measure the reduction in pollution and the preservation of natural resources as a result of effective waste management. It provides global adoption, where this technology becomes a standard in waste management practices across countries and municipalities. Mobile applications will become an integral part of waste management systems, enabling users to report issues, schedule pickups, and get real-time information on waste collection and sorting. Smart waste sorting systems will be integrated into the broader concept of smart cities.

X. REFERENCES

1. Jiu Huang, Pretz, T., Zhengfu Bian, "Intelligent solid waste processing using optical sensor-based sorting technology", in: Proceedings of 3rd IEEE International Conference on Image and Signal Processing (CISP), Vol.4, pp.1657-1661, October 2013, 16-18.
2. Ohtani, K.; Baba, M., "A Simple Identification Method for Object Shapes and Materials Using an Ultrasonic Sensor Array", in Proceedings of the IEEE on Instrumentation and Measurement Technology Conference, IMTC, pp. 2138 - 2143, April 2006, 24-27.
3. Suwon Shin, Kaiyuan Fan, "Smart Automatic Recycling Trash Basket", 2012.
4. Madan Kumar, L., Pavan, B., Kalyan, P.V. Paul, N.S., "Design of an embedded based control system for efficient sorting of waste plastics using Near Infrared Spectroscopy", in: Proceedings of IEEE International Conference on Electronics, Computing and Communication Technologies (IEEE CONNECT), January 2014, pp. 1-6
5. Yann Glouche, Paul Couderc, "A Smart Waste Management with Self-Describing Objects", in Proceedings of Leister, Wolfgang and Jeung, Hoyoung and Koskelainen, Petri. The Second International Conference on Smart Systems, Devices and Technologies (SMART'13), June 2013, Rome, Italy. 2013.
6. S. Sudha, M. Vidhyalaskhmi, K. Pavithra, K. Sangeetha and V. Swathi, "An automatic classification method for environment: Friendly waste segregation using deep learning," IEEE Technological Innovations in ICT for Agriculture and Rural Development (TIAR), pp. 65-70, 2016.

7. N. Sivakumar, A. R. Kunwar, S. Patel, S. Kumar, and S. P. Mala, "Design and development of an automatic clustered, assorted trash segregation system," IEEE International Conference on Recent Trends in Electronics, Information & Communication Technology (RTEICT), pp. 409-413, 2016.
8. M. O. Rahman, A. Hussain, E. Scavino, H. Basri and M. A. Hannan, "Intelligent computer vision system for segregating recyclable waste papers," Expert Systems with Applications, vol. 38, no. 8, pp. 10398- 10407, 2011.
9. T. J. Lukk, T. Tossavainen, J. V. Kujala and T. Raiko, "ZenRobotics Recycler-Robotic sorting using machine learning," Proceedings of the International Conference on Sensor-Based Sorting (SBS), 2014.
10. A. Shaukat, Y. Gao, J. A. Kuo, B. A. Bowen, and P. E. Mort, "Visual classification of waste material for nuclear decommissioning," Robotics and Autonomous Systems, vol. 75, pp. 365-378, 2016.