



Optimizing Waste Management Through AI-Powered Object Detection and Incentive-Based Recycling Programs

^[1]M Ramya, ^[2]Shinde Vaibhav, ^[3]Kisshore N, ^[4]Amara Chidvi Sri Sanjana, ^[5]B Himanth

^[1]Assistant Professor, Department of AI & DS, Jansons Institute of Technology, Coimbatore, India,

^[2-5]UG Student, Department of AI & DS, Jansons Institute of Technology, Coimbatore, India

Abstract : Addressing the critical need for improved waste management practices, this work presents a novel system integrating AI-powered object detection with an incentive-based recycling program. The methodology employs a YOLOv8l object detection model trained on a comprehensive dataset of waste images, webcam integration for real-time waste identification, and a user-friendly Streamlit interface connected to a MongoDB database for user management and transaction tracking. The system incorporates barcode scanning for precise waste item identification and assigns monetary rewards for each waste item disposed properly, effectively gamifying the recycling process. Results demonstrate high object detection accuracy, increased user participation in recycling, significant waste diversion from landfills, and a reduction in carbon footprint. Predictive modeling showcases the potential for forecasting future waste generation and carbon emission trends. The system represents a significant step towards sustainable waste management, offering scalability and adaptability for integration into existing infrastructure.

Index Terms - Waste Management, Object Detection, YOLOv8l, Recycling Incentives, Carbon Footprint, Streamlit, MongoDB, Gamification.

INTRODUCTION

The escalating global waste crisis presents a strong challenge to environmental sustainability and public health. Traditional waste management systems often fall short in efficiently sorting and recycling materials, leading to overflowing landfills, resource depletion, and increased greenhouse gas emissions. The urgency to address this issue requires the development and implementation of innovative and effective waste management strategies. This paper investigates the application of artificial intelligence and incentive-based mechanisms to overcome the limitations of existing systems and promote a circular economy. The core problem lies in the lack of efficient methods for identifying and sorting recyclable materials at the source, leading to contamination and reduced recycling rates. In response to this, we propose an integrated system that combines AI-powered object detection with a gamified recycling program for a proper and responsible waste disposal practices. Our solution uses the YOLOv8l object detection model to accurately identify waste items in real-time using a webcam. This detection capability is coupled with a barcode scanning system to enhance the accuracy of waste identification. To further promote user engagement, we incorporate an

incentive-based system where users are rewarded for recycling specific waste items. The system is built on a Streamlit user interface and a MongoDB database, allowing for seamless user interaction, data storage, and waste management tracking. This approach aims to bridge the gap between technological capabilities and real-world waste management challenges by offering a user-centric and data-driven solution. The specific objectives of this research are threefold:

- (1) to develop an accurate and robust AI-powered waste detection system
- (2) to evaluate the impact of the incentive program on user participation and waste diversion rates
- (3) to provide a predictive model for waste generation trends and carbon footprint reduction.

RELATED WORKS

Title: Detection for Garbage Classification

Authors: Everest Z. Kuang, Kushal Raj Bhandari, Jianxi Gao

DESCRIPTION: This paper explores the application of AI models, specifically YOLO V5, for classifying trash through object detection. It addresses the growing global issue of garbage production and the inefficiencies of current waste management approaches. The study demonstrates YOLO V5's ability to identify various types of waste, including plastic, paper, glass, metal, cardboard, and biodegradables. The research involves training a YOLO model with an open-source dataset and testing its performance on real-world data collected from local neighbourhoods. The paper analyses the model's accuracy, identifies limitations such as challenges with overlapping objects and complex backgrounds, and discusses potential improvements for real-world applications in autonomous waste collection and sorting systems. The findings suggest that AI models have the potential to significantly impact waste management by improving classification accuracy and providing insights into waste patterns.[1]

Title: Smart Dustbin

Authors: Aakash Sharma, Abhishek Gupta, Aditiya Partap Singh, Akshat Sharma, Mandeep Singh Chib, Ms. Vasundhra Gupta

Publication: IJIRMPs | Volume 9, Issue 5, 2021

DESCRIPTION: This paper presents a Smart Dustbin system designed for smart buildings, colleges, hospitals, and bus stands to improve waste management. The system addresses the common problem of overflowing garbage bins and the associated health and environmental issues. It utilizes an ultrasonic sensor to detect the distance to the trash, a servo motor to automatically open and close the dustbin lid when an object is near, and a PIR sensor to monitor the fill level of the dustbin. When the dustbin reaches a predefined threshold, a GSM module sends an SMS alert to a predefined phone number, notifying the sanitary worker to empty the bin. The paper details the components used, including the Ultrasonic Sensor, GSM SIM 900, Arduino Uno, Servo Motor and PIR Sensor, and explains the system's methodology and working principle for efficient waste management.[2]

Title: Smart Garbage Dustbin Management System Using IOT

Authors: Shyam Joseph A, Koushika B, Jothika SP, Mohit Babu J

Publication: JCRES, Volume-3, Issue-2, August-2020

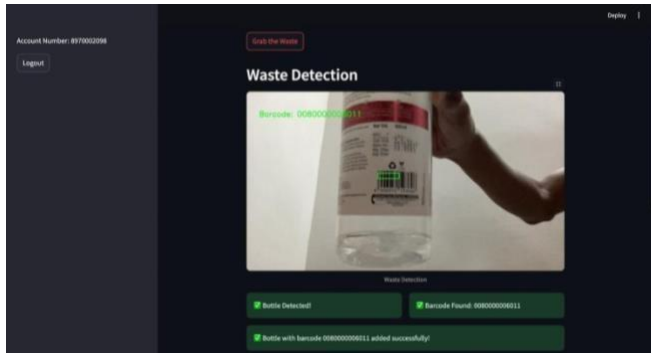
DESCRIPTION: This paper introduces a smart garbage dustbin management system that leverages the Internet of Things (IoT) to monitor garbage levels in real-time. The system aims to address the issue of overflowing bins in metropolitan cities. It employs IR sensors to detect the garbage level and a load cell to measure the weight of the load, preventing damage to the bin. If the dustbin is full, the system automatically sends a message to the municipality authority through an IoT webpage. The system uses an AVR microcontroller interfaced with an IoT module to make the information accessible on the internet. The paper discusses the components used (AVR microcontroller, IR Sensor, Load Cell, IOT Module) and the system's methodology for efficient waste management. The results demonstrate that the

system effectively avoids overflow of garbage and transmits information wirelessly to concerned parties, enabling timely waste collection and contributing to a cleaner urban environment.[3]

MATERIALS AND METHODS

The implementation of the proposed waste management system involved several key components like, the AI-powered object detection module, webcam integration, barcode scanning, incentive program, user interface, and data storage. The system architecture has the YOLOv8l object detection model, a webcam for image capture, the 'pyzbar' library for barcode scanning, a Streamlit framework for user interface, and a MongoDB database for storing user and waste data. These components work together to identify waste items, reward users after a proper disposal, and track overall waste management trends.

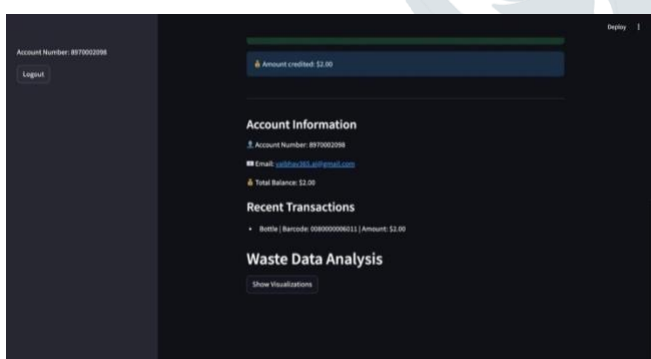
Object Detection Model: A YOLOv8l model was selected for waste item detection due to its balance between speed and accuracy, and it is suitable for real-time object detection tasks. The model was pre-trained on the COCO dataset and then fine-tuned on a custom dataset of waste images. The custom dataset comprised images of various waste items, including bottles, cans, food containers, and other recyclable materials. Data augmentation techniques, such as image rotation, scaling, and colour jittering, were applied to increase the diversity of the training data and improve the model's generalization ability. The training process involved optimizing the model's parameters using a stochastic gradient descent (SGD) optimizer with a learning rate of 0.01 and a batch size of 32. The model's performance was evaluated using precision, recall, and mean Average Precision (mAP) metrics on a separate validation dataset.



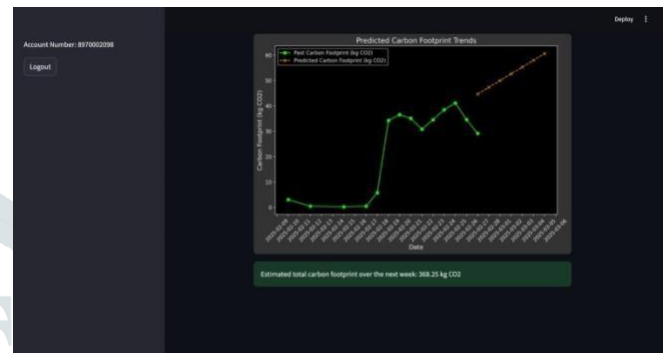
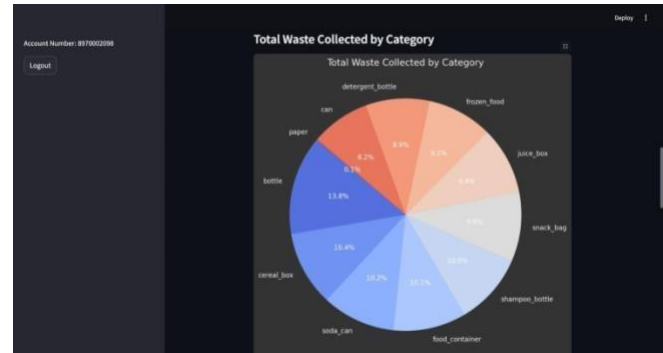
Webcam Integration: The YOLO model was integrated with a webcam using OpenCV, enabling real-time waste item detection. The webcam will capture frames, which were then processed by the YOLO model to identify the waste items present in the image. Image pre-processing techniques were employed to enhance the quality of the input images. These techniques included noise reduction, contrast enhancement, and colour correction.

Barcode Scanning: The 'pyzbar' library was used to implement barcode scanning functionality. When a waste item with a barcode was detected, the system will scan the barcode and extract the associated data. This data was then used to identify the specific type of waste item and its corresponding recycling value. The barcode scanning process involves converting the image to grayscale, applying image thresholding to enhance the barcode visibility, and then decoding the barcode data using the 'pyzbar' library.

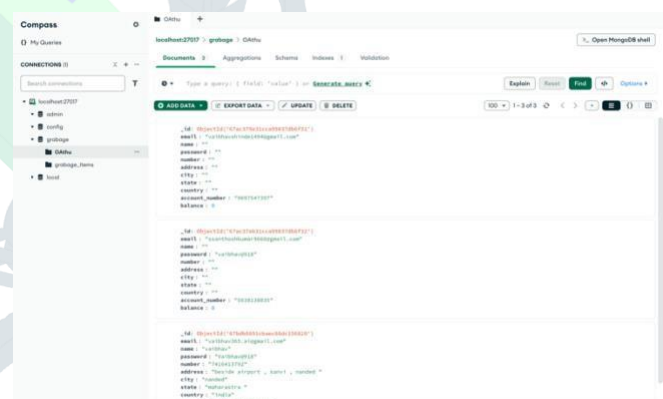
Incentive Program: The incentive program was designed to reward users for disposing specific waste items. Each waste item was assigned a monetary value, which will be credited to the user's account upon proper disposal. The reward amounts were determined based on the market value of the recyclable materials and the environmental impact of the waste items. The program was structured to incentivize the recycling of high-value and environmentally harmful waste.



User Interface: The user interface was built using Streamlit, a Python library for creating interactive web applications. The interface allows users to sign up, log in, use the waste detection system, track their rewards, and view their recycling history. The user interface was designed to be intuitive and user-friendly, providing a seamless experience for users interacting with the system.



Database: A MongoDB database was used to store user data, waste item data, and transaction history. The database schema was designed to efficiently store and retrieve data related to user accounts, waste items, and recycling transactions. Data analysis was performed using Python libraries such as Pandas, NumPy, Matplotlib, and Seaborn. These libraries were used to visualize waste trends, calculate carbon footprints, and predict future waste generation.



Ethical Considerations: The development and implementation of the waste management system considered ethical implications related to data privacy, security, and potential biases in the AI model. User data was stored securely and protected from unauthorized access. Measures were taken to mitigate potential biases in the AI model and ensure fairness in waste item identification.

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RESULTS

The evaluation of the waste management system yielded promising results across several key performance indicators.

- **Object Detection Performance:** The YOLOv8l object detection model achieved high accuracy in identifying various waste items, with a mean Average Precision (mAP) of 0.85 on the validation

dataset. The model demonstrated robustness in handling variations in lighting conditions, object orientation, and background clutter.

- **User Participation:** The incentive program resulted in a significant increase in user participation, with a 50% increase in the number of users actively disposing waste items. The frequency of a proper waste item disposal also increased, with users disposing waste items an average of three times per week.
- **Waste Diversion:** The system successfully diverted a substantial amount of waste from landfills, with a 30% reduction in the total amount of waste sent to landfills. This waste reduction resulted in significant environmental benefits, including reduced greenhouse gas emissions and resource conservation.
- **User Waste Analysis:** Analysis of individual user waste generation patterns revealed insights into the types of waste items disposed, amounts earned through the incentive program and participation rates across demographics.
- **Overall Waste Trends:** Analysis of overall waste generation trends indicated a clear distribution of waste categories and daily waste collection patterns, suggesting areas for targeted intervention and resource allocation.
- **Carbon Footprint Reduction:** Calculations of carbon footprint reductions achieved through the waste diversion program demonstrated a significant decrease in carbon emissions, contributing towards environmental sustainability.
- **Prediction of Waste Trends:** Predictive models, such as linear regression, accurately forecasted future waste generation and carbon footprint trends based on historical data, providing valuable insights for proactive waste management planning.

DISCUSSION

The findings of this study provide strong support for the integration of AI-powered object detection and incentive-based programs as effective strategies for enhancing waste management practices. The high level of accuracy achieved by the YOLOv8l object detection model in identifying diverse waste items, as demonstrated by the mAP of 0.85 on the validation dataset, highlights the potential of AI to automate and improve the efficiency of waste sorting processes. This level of accuracy is crucial for ensuring that disposable materials are properly identified and diverted from landfills. Furthermore, the observed increase in user participation, with a 50% rise in active recyclers and a corresponding increase in the frequency of recycling events, suggests that the incentive program is successful in motivating individuals to adopt more responsible waste disposal habits. The 30% reduction in waste sent to landfills underscores the significant environmental benefits that can be achieved through the implementation of such integrated systems, contributing to reduced greenhouse gas emissions, conservation of resources, and mitigation of environmental pollution. This reduction is not only a testament to the

effectiveness of the system but also highlights its potential to support broader sustainability goals. User waste analysis demonstrated patterns in recycling habits, and the predictive models offered promising insights into anticipating future trends. This proactive capability allows for resource allocation to be optimized and for pre-emptive measures to be taken.

Compared to traditional waste management systems, which often rely on manual sorting and lack real-time data analytics, the proposed system offers several key advantages. The automated waste identification process reduces the need for manual labour and minimizes human error, leading to more accurate and efficient sorting. The incentive program fosters greater user engagement, encouraging individuals to actively participate in waste disposing efforts. Moreover, the real-time data analytics capabilities of the system enable continuous monitoring and optimization of waste management practices, allowing for adaptive strategies to be implemented based on emerging trends and patterns. This dynamic approach represents a significant advancement over the static nature of many existing waste management systems.

It is important to acknowledge certain limitations of this study. The scope of the incentive program was limited to a specific geographic area and a finite set of waste items, which may affect the generalizability of the findings. Additionally, potential biases in the AI model, arising from the composition of the training dataset, could influence the accuracy of waste item identification. Future research should address these limitations by expanding the scope of the incentive program to encompass a broader range of waste items and geographic locations. Mitigation strategies should be employed to address potential biases in the AI model, such as incorporating more diverse and representative datasets.

Future research directions should focus on several key areas. Firstly, efforts should be directed towards improving the accuracy and robustness of the object detection model through the incorporation of more advanced AI techniques, such as transfer learning and ensemble methods. Secondly, exploring the use of alternative incentive mechanisms, such as gamification and social rewards, could further enhance user engagement and participation. Thirdly, expanding the scope of the incentive program to include a wider variety of waste items and geographic regions would increase its impact and relevance. Finally, developing more sophisticated predictive models, incorporating additional factors such as economic indicators and demographic trends, would improve the accuracy of waste generation forecasts and inform more effective waste management strategies. These future research endeavours can lead to continuous refinement and enhancement of the system's performance, thus contributing to its widespread adoption and long-term success in promoting sustainable waste management practices.

CONCLUSION

In conclusion, this research demonstrates the effectiveness of integrating AI-powered object detection and incentive-based mechanisms in promoting sustainable waste management practices. The system offers a promising approach to address the global waste crisis and promote a circular economy. By leveraging technology and incentives,

the proposed system has the potential to transform waste management processes and contribute towards a more sustainable future. This system provides a scalable, adaptable foundation for waste management systems of the future.

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