*Research Article*

Smart City Waste Management Using Text Data and Images

Jaya Srinivasa Sailesh 1, Atam Vidya Manoj Elle 2 and Rajesh 2,\*

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1. Affiliation 1; [jgaddam@saintpeters.edu](mailto:jgaddam@saintpeters.edu)
2. Affiliation 2; [aelle@saintpeters.edu](mailto:aelle@saintpeters.edu)
3. Affiliation 3; [rajesh@saintpeters.edu](mailto:rajesh@saintpeters.edu)

**\*** Correspondence: e-mail@e-mail.com; Tel.: (optional; include country code; if there are multiple corresponding authors, add author initials)

**Featured Application: Authors are encouraged to provide a concise description of the specific application or a potential application of the work. This section is not mandatory.**

**Abstract:** This study focuses on integrating text data and images to enhance waste management in smart cities. We employ advanced machine learning and computer vision techniques to automate waste sorting. Using the TACO dataset, we utilize convolutional neural networks (CNNs) for image-based waste identification and natural language processing (NLP) models for analyzing text data. By combining these modalities, our system aims to improve the accuracy of waste sorting, minimize manual errors, and optimize waste collection processes, ultimately contributing to a more sustainable and efficient urban environment.

**Keywords:** Smart City, Waste Management, Machine Learning, Image Classification, IoT, Reinforcement Learning

1. Introduction

The introduction sets the stage by discussing the significance of smart city technologies in managing waste. It highlights the challenges faced by current waste management systems, such as slow processes and human errors. The goal is to explain how using text data and images with machine learning can improve waste management. This approach helps reduce mistakes, enhance efficiency, and contribute to a more sustainable environment.

2. Materials and Methods  
  
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**Data Collection**: We use the TACO dataset, which contains labeled images of waste from different environments. These images are supplemented with text data from smart sensors and databases to provide a complete view of the waste management needs.

**Model Selection**: We chose models like YOLO for image classification and BERT for text processing because they are well-suited to handle the specific types of data in our project.

**Data Preprocessing**: For images, we apply augmentation techniques and normalize them. Text data is processed through tokenization and embedding to convert it into a usable format.

**Integration Strategy**: We developed methods to combine the outputs from the image and text models. This involves feature fusion and decision-level integration to create a unified system for waste management.

**Deployment**: The models are deployed on cloud platforms like AWS SageMaker, allowing for real-time processing and management of waste.

3. Results

This section may be divided by subheadings. It should provide a concise and precise description of the experimental results, their interpretation, as well as the experimental conclusions that can be drawn.

* 1. Subsection

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**3.1. Implementation Summary**

1. **YOLO (You Only Look Once)**:  
   • Pre-trained YOLOv5 model for real-time object detection and classification;  
   • Detects multiple objects per image with bounding boxes and labels;  
   • Suitable for tasks requiring quick and multi-object detection.
2. **ResNet50**:  
   • Pre-trained ResNet50 model for single-object classification;  
   • Processes resized images (224x224) and predicts the top-1 label;  
   • Ideal for identifying the primary object in an image.
3. **CNN (Convolutional Neural Network)**:  
   • Custom CNN architecture for image classification;  
   • Optimized for smaller, task-specific datasets;  
   • Provides a balance between speed and accuracy.

**3.2. Results**

1. **YOLO Results**:
   1. Detected multiple objects per image, achieving real-time performance;
   2. Accuracy ranged between **78–85%** on smart city waste images;
   3. Sample results: "plastic bottle", "cardboard box", "trash bin".
2. **ResNet50 Results**:
   1. Achieved **75–80% accuracy** for single-object classification;
   2. Efficiently processed images resized to 224x224;
   3. Sample results: "bottle", "can", "paper", "plastic bag".
3. **CNN Results**:
   1. Custom CNN achieved **70–75% accuracy** on waste classification tasks;
   2. Faster inference for small, task-specific datasets;
   3. Sample results: "glass bottle", "metal can", "plastic wrapper".

3.2. Figures, Tables and Schemes

All figures and tables should be cited in the main text as Figure 1, Table 1, etc.

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**Figure 1.** This is a figure shows model ResNet50

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**Figure 2.** This is a figure shows model CNN

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**Figure 3.** This is a figure shows model YOLOv5

**Table 1.** This is a table shows the different classification model accuracy, speed and Purpose.

|  |  |  |  |
| --- | --- | --- | --- |
| **Model** | **Accuracy** | **Speed** | **Purpose** |
| YOLOv5 | 78–85% | Real-time | Detect and classify multiple objects |
| ResNet50 | 75-80% | Moderate | Single-object waste classification |
| CNN | 70-75% | Fast | Task-specific Classification |

1

Tables may have a footer.

3.3. Formatting of Theorems

**Neural Networks (CNNs) for Image Classification**

**Theorem 1**: In a waste classification system using Convolutional Neural Networks (CNNs), the model accuracy improves significantly with increased dataset size and data augmentation techniques.

**YOLO for Real-Time Object Detection**

**Theorem 2**: YOLO (You Only Look Once) outperforms traditional object detection models in real-time applications by processing images in a single pass, achieving faster inference speeds with acceptable accuracy in waste detection scenarios.

**ResNet50 for Image Classification and Transfer Learning**

**Theorem 3**: ResNet50, when fine-tuned on domain-specific datasets (e.g., waste classification), can achieve high classification accuracy even with a relatively small amount of labeled data by leveraging pre-trained weights from general image classification tasks.

4. Discussion

In this study, we applied machine learning models (ResNet50, YOLO, and CNN) to waste management in smart cities. Each model had its own strengths, and using them can really improve how waste is classified and managed.

* **ResNet50 Results**

ResNet50, which was fine-tuned for waste classification, performed well, achieving around **75-80% accuracy**. This matches previous studies showing that pre-trained models like ResNet50 can adapt well to new tasks with smaller datasets. To improve accuracy, future work could focus on using larger datasets or more specific features.

* **YOLO Results**

YOLO’s real-time object detection allowed it to identify multiple objects in images, with an accuracy of **78-85%**. This is essential for applications like managing waste bins in smart cities, where speed is important. YOLO’s efficiency and speed make it ideal for real-time tasks. Future research could look into integrating YOLO with IoT-enabled waste bins for fully automated waste detection and classification.

* **CNN Results**

The custom CNN model achieved an accuracy of **70-75%**, which is suitable for smaller datasets. While CNNs didn’t perform as well as YOLO or ResNet50, they are useful when resources are limited or when custom architectures are needed. Future work could explore hybrid models that combine CNNs with other networks for improved performance.

5. Conclusions

Results show that machine learning models, especially those trained on large datasets, can be powerful tools for waste management. **ResNet50** and **YOLO** outperformed traditional CNN models in complex tasks like waste classification and detection.

These findings show that using machine learning models with IoT sensors in smart cities could make waste management faster and more sustainable. Future work should focus on combining real-time data from sensors with waste classification models, improving the training data, and trying different models together to get the best results.

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    School of Computer Science and Engineering, Vellore Institute of Technology, Chennai, Tamilnadu, and Department of Electronics and Communication Engineering, Koneru Lakshmaiah Education Foundation, Vijayawada, Andhra Pradesh, India.  
    Emails: gayathri.r@vit.ac.in, sholausha.rani@vit.ac.in, christie.vincent2021@vitstudent.ac.in.