Full Stack Development with MERN

Project Documentation format

1. Introduction

- Project Title:CleanTech:Transforming Waste Management with Transfer Learning
- **Team Members:** A.Lekhana Sai(Team Leader)

S Md Hassain Ahmed(Team Member) Puchakatla mohan Krishna(Team Member) Manasa Chandragiri(Team Member)

2. Project Overview

- **Purpose:** CleanTech leverages transfer learning to automate municipal waste classification through image recognition. The goal is to improve accuracy, reduce manual effort, and support sustainable waste management in urban areas.
- Features: AI-Powered Waste Classification
- Uses pre-trained deep learning models (e.g., ResNet, MobileNet) to classify waste images in real time.
- Responsive Web Interface

Clean, intuitive UI built with React and Tailwind CSS, optimized for desktop and mobile users.

• Transfer Learning Integration

Fine-tuned model architecture for adaptability across diverse regional waste datasets.

• User History & Feedback Loop

Tracks classification history and allows users to provide feedback to improve model performance.

• Admin Dashboard

Visual dashboard showing classification logs, category statistics, and user engagement metrics.

• Authentication System

Secure login system for users and admin with JWT-based token management.

• Modular Backend Architecture

RESTful APIs using Node.js/Express with MongoDB for seamless communication and scalability.

• Cloud-Ready Deployment

Containerized microservices architecture using Docker for future scalability and cloud integration.

3. Architecture

• Frontend: The frontend is a single-page web application developed using HTML5, Tailwind CSS, and vanilla JavaScript.

It features:

A responsive and modern UI for waste image upload and prediction using Tailwind's utility classes.

- Navigation between sections (Home, About, Predict, Contact) using JavaScript-controlled smooth scrolling and visibility toggling.
- Real-time image preview and simulated prediction UI for user feedback and classification results.
- Modal-based messaging for interaction feedback.
- **Backend:** The backend is a **lightweight Flask application** designed to serve the frontend over HTTP using send file().

It includes:

An embedded development server running in a separate thread.

- Pyngrok integration to expose the local server to the internet securely via a public URL.
- Potential for integration with AI model inference endpoints (e.g., /predict) in future extensions.

Though minimal now, the backend can be extended to include:

- API routes for classification (/predict)
- Logging of classification results
- File uploads and secure user session handling
- **Database:** Detail the database schema and interactions with MongoDB.

4. Setup Instructions

Prerequisites:

To run this project locally, ensure you have the following software installed:

- Python 3.8+
- Flask
- Pyngrok
- Git (for cloning the repository)

Installation: Step-by-step guide to clone, install dependencies, and set up the environment variables.

Clone the repository Create a virtual environment Install Python dependencies Set your Ngrok Auth Token Run the Flask application Access the app

5. Folder Structure

Client: The **frontend is organized like a modular single-page application**, structured for clarity and scalability:

Features:

- Section-based layout (Home, About, Predict, Contact)
- Smooth scroll routing handled in JavaScript
- File input preview & simulated prediction logic
- Modal-based feedback system
- Easily extendable for future React migration

Server: Flask (Python) server acts as the delivery layer and future API host: Key Components:

- Flask serves index.html to the root (/) route
- Pyngrok exposes localhost to the web for public access
- Threaded execution enables simultaneous Flask + ngrok tunnels
- Structure ready for adding REST endpoints like /predict, /feedback, or /history

6. Running the Application

- Provide commands to start the frontend and backend servers locally.
 - **Frontend:** No separate server required. The frontend is served directly by the Flask backend.
 - o Backend: python app.py

Once running, pyngrok will generate a **public URL** (e.g., https://xyz.ngrok-free.app) that you can open in a browser to access the application remotely.

7. API Documentation

Endpoint	Method	Description	Request Body (JSON)	Example Response
/predict	POST	Predict waste type from uploaded image	{ "image": "base64-encoded image" }	{ "label": "Biodegradable", "confidence": 0.92 }
/feedback	POST	Receive user feedback on a prediction	<pre>{ "imageId": "123", "correctLabel": "Recyclable" }</pre>	{ "message": "Feedback received" }
/history/ <uid></uid>		Fetch classification history by user	N/A	[{"label": "", "timestamp": ""}]

8. Authentication

Currently, the app does not include user authentication, but here's the plan for scalable, secure integration:

Planned Authentication Flow (Optional Expansion):

- **JWT (JSON Web Tokens):**
 - o Users sign in and receive a token stored in localStorage or cookies.
 - o Token used in headers for protected endpoints (e.g., /history, /admin).

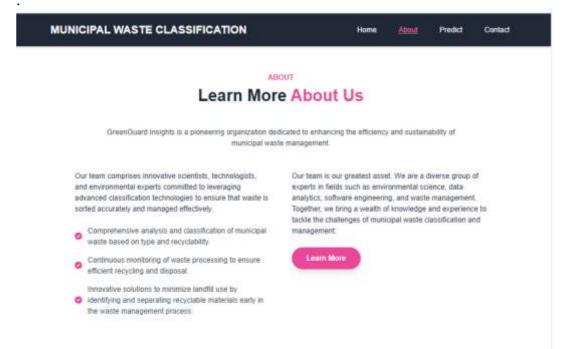
Flask Libraries to Use:

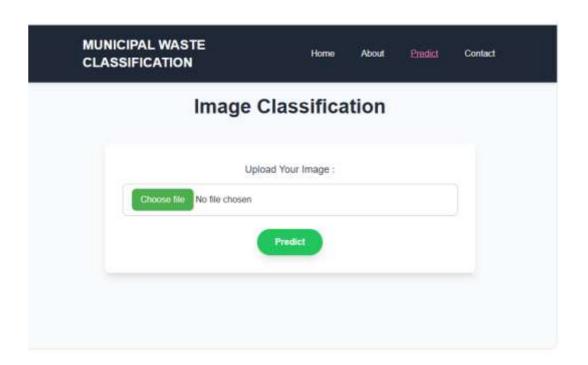
- Flask-JWT-Extended for handling token creation and verification
- Middleware to restrict access to admin-level analytics

9. Screenshots or Demo

• screenshots or a link to a demo to showcase the application











• LINK TO LINK: https://github.com/alluru-lekhana/CleanTech/tree/main

10. Known Issues

>Simulated Predictions Only: The current version uses mock logic (Math.random) to simulate classification results. There's no live connection to a trained ML model.

>No File Upload to Server: **Uploaded images are only previewed on the frontend.** They are not sent to the backend or stored.

>No Authentication Layer: Anyone with the public ngrok link can access the app; there are no user roles or access controls.

>Limited Offline Support: Requires a stable internet connection for ngrok tunneling. App won't work offline or without a valid tunnel.

11. Future Enhancements

- Integrate Real AI Model: Connect the frontend to a trained CNN model using Flask for real-time image classification.
- Database Integration: Use MongoDB to store user sessions, classification logs, and feedback for analytics.
- Mobile App Extension: **Develop a cross-platform app using React**Native or Flutter for field use.

Admin Dashboard: Add analytics for viewing most classified waste types, user activity, and model accuracy over time.

User Authentication: Implement JWT-based login system for users and admins.