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Group Number: 3A

Mentor – Dr. Niranjan Deshpande

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Links to Codebases

- Yashodip More: [GitHub link To be added]
- Komal Kumavat: [GitHub link To be added]
- S.M. Sakthivel: [GitHub link To be added]
- Barun Saha: [GitHub link To be added]
- Bibaswan Das: [GitHub link To be added]

Selected Domain & Problem Statement:

Domain: Agriculture & Environment

Problem Statement:

"AgriWaste2Fuel - Smart Waste Converter with GHG & Carbon Credit Tracking"

To build an AI-powered platform that helps farmers detect and classify farm waste, guides them in converting it into compost or biogas, and estimates the environmental benefits in terms of greenhouse gas (GHG) emission savings and carbon credits turning farm waste into both sustainable energy and income.

Module 1: Farm Waste Detection & Classification

Goal:

Identify the type of farm waste using either **image-based AI detection** or **manual text input**, to determine whether it should be composted or processed for biogas.

Option 1: Image-Based Detection

- 1. Farmer uploads a photo of the waste using a smartphone or device.
- 2. YOLOv8 model detects and classifies waste into types like:
 - Cow dung
 - Paddy straw
 - Vegetable/kitchen scraps
 - Leaf litter, etc.
- 3. Model also estimates **waste quantity** based on image reference or asks the farmer to enter it.

Option 2: Manual Input

- 1. **Farmer types the name** of the waste (e.g., "cow dung", "banana peels", "sugarcane bagasse").
- 2. System **maps it to a known waste category** using a backend dictionary or NLP-based keyword matcher.
- 3. The input is validated for **spelling & matching category** (basic NLP or regex).

Output of This Module:

- Waste Type (image/manual)
- Quantity (auto-estimated or manually entered)
- Confidence score (if image-based)
- Waste classification label → passed to recommendation engine

Technologies Used:

- YOLOv8 (for object detection from images)
- Python FastAPI (backend logic)
- **NLP tools** (for validating manual input)
- **Frontend**: Form inputs + image upload (React/Flutter)

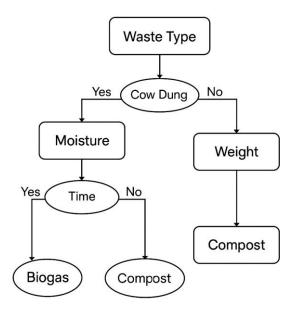
Module 2: Biogas or Compost Recommendation Engine

Goal:

Recommend the most suitable waste conversion process: **biogas generation** or **composting**, based on waste characteristics and quantity.

- 1. Decision tree logic evaluates:
 - o High-moisture organic waste (e.g., cow dung + wet food waste) → Biogas
 - o Dry waste (e.g., crop residues + leaf litter) \rightarrow Compost
- 2. Output includes:
 - Suggested method (biogas or compost)
 - Steps to process it
 - Tools/materials required
 - Expected output (biogas in m³ or compost in kg)

Example:



Module 3: GHG Savings Prediction

Goal:

Estimate how much CO₂e emissions are avoided by converting waste instead of burning it.

Procedure:

- 1. Use the classified waste type and approximate weight.
- 2. Refer to standard emission factors (e.g., from IPCC, FAO).
- 3. Apply a simple formula: GHG saved = weight \times (EF burning EF conversion)
- 4. Output the GHG savings in tons of CO2e. E_F-Emission Factor

Module 4: Carbon Credit Estimation

Goal:

Convert GHG savings into carbon credits and show the farmer how much they can potentially earn.

Procedure:

- 1. 1 ton of CO_2e saved = 1 carbon credit.
- 2. Multiply credits by market rate.
- 3. Show credits earned and estimated income.
- 4. Optionally generate a digital certificate and connect with verified platforms for credit sale.

Detailed Tech Stack

1. Frontend (User Interface)

Component	Tech
Mobile/Web App	Flutter (cross-platform) OR ReactJS
Form Inputs	Image upload, text fields
Display	Result Dashboard (waste type, GHG saved, credits)

2. Backend (Logic + APIs)

Component	Tech
API Framework	FastAPI (Python, async-ready)
Image Handling	OpenCV, Pillow
Waste Matching (Manual)	NLTK or regex + dictionary
Rule-based Engine	Custom logic or Decision Trees
GHG Calculator	Python logic using IPCC data
Credit Estimator	Python calculator module

3. AI/ML Models

Purpose	Tool/Framework
Waste Detection	YOLOv8 (Ultralytics, PyTorch)
Waste Type Classifier	Fine-tuned if needed
Model Hosting	TorchServe, Flask API, or FastAPI route

4. Database

Data Type	DB Engine
Waste & GHG Factors	PostgreSQL / SQLite
Carbon Credit Prices	API or local JSON DB
User Profiles, Input History	Firebase / Supabase / MongoDB

5. Deployment & Infra

Purpose	Tool
Cloud Hosting	Render, Railway, Vercel, Heroku (dev)
Production Backend	AWS EC2 / Azure App Service
Model Storage	GitHub, HuggingFace, S3
Certificates & Emails	SMTP + PDFkit / ReportLab (Python)

Weekly Activities & Progress Summary

Team Participation & Approach

- The team started by discussing the **real on-ground issues** faced by farmers in Maharashtra regarding waste handling.
- We explored multiple ideas, but concluded that the main value is in fuel (biogas) and compost generation, and designed our system around that.

Technical Work Done

- Created the initial architecture and flowchart for biogas/compost decision engine.
- Designed a **decision tree** model to determine best treatment method.
- Identified features for **yield prediction models** (waste type, quantity, temperature, moisture).
- Explored dataset sources: ICAR, MNRE, TERI, research papers.
- Created image-based decision tree diagrams and project visual flow.
- Discussed integrating GFG guidelines and carbon offset tracker as additional layers.

Obstacles Faced

- Finding real, open-source datasets on compost/biogas yield was challenging.
- Defining a clear balance between **core functionality** (waste to fuel) and **add-ons** (carbon tracking, government guidelines).
- Aligning the team around a **focused direction**, especially after adding multiple modules.

Guidance from Mentor

- Mentor helped us **re-center our focus** on the core problem of fuel/compost generation, not just data dashboards.
- Suggested to prioritize farmer usability and real-world application over just ML for the sake of ML.
- Guided us in structuring the problem clearly and recommended looking into **simple** regression models for early yield estimation.