

# Machine Learning Project Documentation

## format

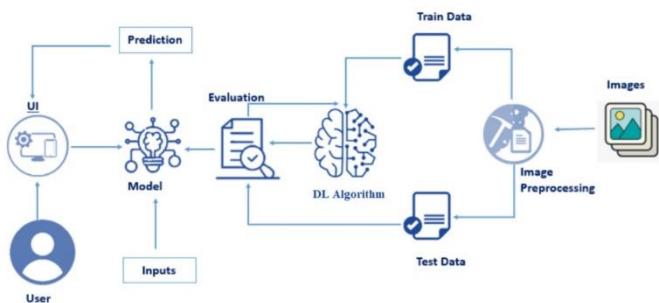
### 1. Introduction

- **Project Title:** Pollen's Profiling: Automated Classification of Pollen Grains
- **Team Members:**
  1. Eduri Maryjones (Data collection, Train the model, Application building)
  2. Idimukkala Yasasswini (Train the model, save the model, Test the model, Application building)
  3. Mamidela Venkata Naga Suseel Kumar (Read the data, Image pre-processing, Training the model, Save the model, Test the model, Application building)
  4. Inturi Venkata Vikash (Data Collection, Exploratory Data Analysis, Image Pre-processing, Application Building)
  5. J Pushpitha (Data Collection, Exploratory Data Analysis, Image Pre-processing, Application Building)

### 2. Project Overview

- **Purpose:** Automate pollen grain identification using a deep learning CNN model based on image classification.
- **Features:** Image upload for pollen classification.  
Prediction of pollen type.  
Web-based interface using Flask.

### 3. Architecture



### 4. Setup Instructions

- **Prerequisites:** Python, TensorFlow, Flask, OpenCV, Scikit-learn, Numpy, Pandas, Matplotlib.
- **Installation:**  
`git clone https://github.com/EduriMaryJones/Pollen-s-Profiling-Automated-Classification-of-Pollen-Grains.git`  
`cd Pollen-s-Profiling-Automated-Classification-of-Pollen-Grains`

```
cd flask  
pip install -r requirements.txt
```

## 5. Folder Structure

```
pollens_profiling_project_executable_files/  
|   flasks/           # Main Flask app directory  
|   |   app.py          # Flask backend application  
|   |   pollen_classification.ipynb # Model training and evaluation notebook  
|   |   pollen_model.keras    # Trained CNN model  
|   |   labelencoder.pkl    # Label encoder for class decoding  
|   |   requirements.txt    # Python dependencies  
|   |   templates/         # HTML templates  
|   |       index.html  
|   |       about.html  
|   |       prediction.html  
|   |       team.html  
|   |       contact.html  
|   |   static/            # Static files (CSS, images)  
|   |       style.css  
|   |       images/  
|   |   uploads/           # Uploaded images for prediction  
|   pollen_dataset/      # Dataset directory  
|       images/           # Raw pollen grain images  
|       bboxes.csv        # Bounding box data  
|       class_map.csv     # Class mapping file
```

## 6. Running the Application

- > python app.py
- Type this in the terminal and click the link

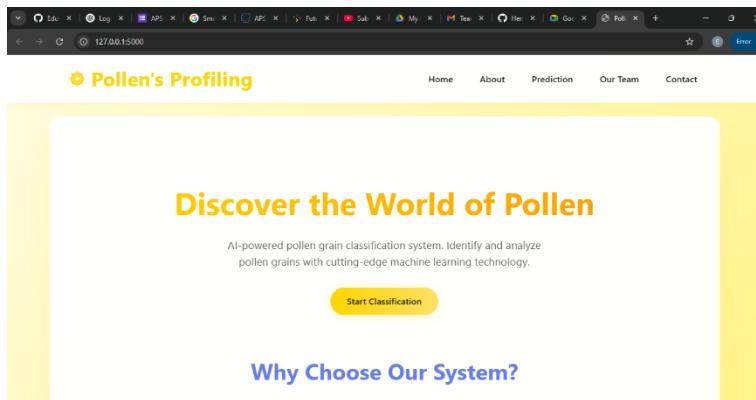
## 7. API Documentation

- Base URL: http://127.0.0.1:5000/
- ► Endpoints:
  - / → Home (GET)
  - /about → About page (GET)
  - /prediction → Prediction page UI (GET)
  - /team → Team page (GET)
  - /contact → Contact page (GET)
  - /predict → Prediction API (POST)
- 🌱 /predict API:
  - Method: POST
  - Input: Image file (file)

## 8. Authentication

- No Authentication required

## 9. User Interface



## 10. Testing

### Testing Strategy

### Manual Testing

- Each feature of the web application (Homepage, Prediction, Upload, Result display, Contact form) was manually tested across different browsers (Chrome, Edge, Firefox).
- File upload validation was tested with:
- Valid image formats (.jpg, .png)
- Invalid inputs (non-image files, empty uploads)

### Model Evaluation

- The CNN model was evaluated using Train/Test Split (80/20) stratified by class

Metrics Used:

- Accuracy
- Confusion Matrix
- Classification Report (Precision, Recall, F1-score)
- Visual performance tracking was done using Matplotlib plots for training/validation loss and accuracy.

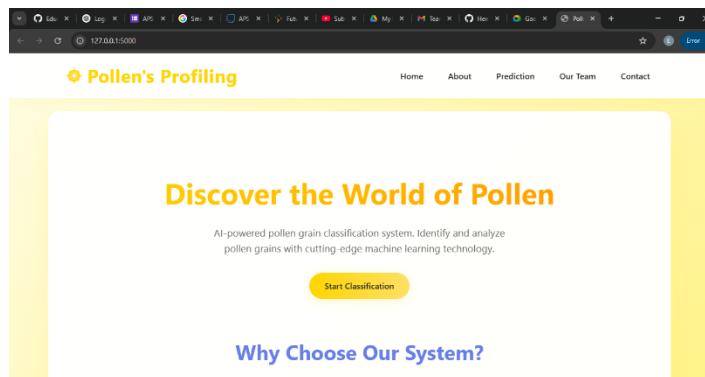
### Prediction Testing

- Real-time predictions were tested by:
- Uploading test images via the /predict endpoint
- Verifying output class labels against known inputs
- Cross-checking consistency with model evaluation results in the notebook

### Tools Used for Testing

- Tool / Library Purpose
- Jupyter Notebook Model training, evaluation, and testing
- Matplotlib Visualization of training metrics
- Scikit-learn Confusion matrix, classification report
- Browser Developer Tools UI inspection and validation
- Flask Debug Mode Monitoring and catching runtime errors

## 11. Screenshots or Demo



## 12. Known Issues

- Accuracy depends on dataset quality.
- Misclassification in very similar pollen types.

## 13. Future Enhancements

- Deploy on cloud (Render, Heroku)
- Add user login to save predictions.
- Improve model with a larger dataset.