# 1. INTRODUCTION

## 1.1 Project Overview

GrainPalette is an AI-powered system that classifies different types of rice grains using image recognition It uses deep learning models and transfer learning techniques to accurately identify rice varieties from ph

## 1.2 Purpose

To develop an efficient, scalable, and automated rice type classification tool that helps farmers, traders, a

#### 2. IDEATION PHASE

### 2.1 Problem Statement

Manual classification of rice types is time-consuming, subjective, and error-prone. There's a need for an automated, intelligent system for accurate grain classification.

# 2.2 Empathy Map Canvas

- Think & Feel: Wants accurate grain analysis
- See: Inconsistent manual sorting
- Say & Do: Uses mobile cameras or lab tools
- Hear: Complaints about quality grading errors
- Pain: Loss due to misclassification
- Gain: Better pricing and trade value

# 2.3 Brainstorming

- Use CNN-based deep learning for image classification
- Apply transfer learning with models like ResNet/VGG
- Build a user-friendly web app for uploading grain images

### 3. REQUIREMENT ANALYSIS

### 3.1 Customer Journey Map

Capture rice image  $\rightarrow$  Upload to app  $\rightarrow$  Model predicts rice type  $\rightarrow$  Display result

### 3.2 Solution Requirement

- Pretrained model (ResNet, MobileNet, etc.)
- Python backend using Flask/FastAPI
- Frontend (React/HTML/CSS/JS)
- Dataset of labeled rice grain images

### 3.3 Data Flow Diagram

 $User \rightarrow Upload Image \rightarrow Backend \rightarrow Deep Learning Model \rightarrow Predicted Type \rightarrow User$ 

## 3.4 Technology Stack

- Frontend: HTML/CSS/JavaScript
- Backend: Python (Flask/FastAPI)
- Al Model: TensorFlow/Keras with transfer learning
- Hosting: Render, GitHub Pages
- Dataset: Rice Image Dataset (e.g., from Kaggle)

#### 4. PROJECT DESIGN

### 4.1 Problem-Solution Fit

GrainPalette addresses the real-world problem of inconsistent grain classification with a fast and accurate

# 4.2 Proposed Solution

A smart system that classifies rice types using a pre-trained deep learning model, fine-tuned on a rice im-

### 4.3 Solution Architecture

Frontend ↔ Backend API ↔ Transfer Learning Model ↔ Result Display

### 5. PROJECT PLANNING & SCHEDULING

- Day 1: Research and dataset collection
- Day 2: Model selection and transfer learning setup
- Day 3: Model training and validation
- Day 4: Build backend and integrate model
- Day 5: Build frontend and test end-to-end
- Day 6: Debugging and performance checks
- Day 7: Documentation and reporting

### 6. FUNCTIONAL AND PERFORMANCE TESTING

### 6.1 Performance Testing

- Accuracy: >90% on validation set
- Latency: Image classification in under 2 seconds
- Confusion Matrix: To check performance per rice type
- Metrics: Accuracy, Precision, Recall, F1-score

### 7. RESULTS

Screenshots of classification results, model metrics like loss and accuracy plots, and sample input/output

### 8. ADVANTAGES & DISADVANTAGES

### Advantages

- Fast and accurate grain classification
- Reduces manual errors
- Helps standardize rice grading

### Disadvantages

- Needs good-quality images
- Depends on training data variety
- Not suitable for mixed-grain images (initial version)

### 9. CONCLUSION

GrainPalette proves how transfer learning can be successfully applied to agriculture. It brings speed, accuracy, and automation to a traditionally manual process.

### 10. FUTURE SCOPE

- Add support for multiple grains (wheat, barley, etc.)
- Develop mobile app version
- Improve accuracy using larger datasets and augmentation
- Integrate with supply chain systems for real-world use

### 11. APPENDIX

- Model Code: Included
- Dataset Source: e.g., Kaggle Rice Dataset
- GitHub Link: github.com/YourRepo/GrainPalette
- Libraries Used: TensorFlow, Keras, OpenCV, Flask