

# Rice Classification – Final Project Report

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## 1 . Project Overview

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This project focuses on rice grain classification using a Convolutional Neural Network model powered by TensorFlow and trained on a dataset of five rice varieties: Arborio, Basmati, Ipsala, Jasmine, and Karacadag. The aim is to automate rice type identification from images to aid quality assurance in agriculture and food industries.

## 1 . Problem Statement

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Manual classification of rice varieties is time-consuming and prone to human error. A deep learning-based classification model can help automate the process and provide consistent and accurate identification, helping producers and regulators maintain quality control.

## 1 . Dataset and Preprocessing

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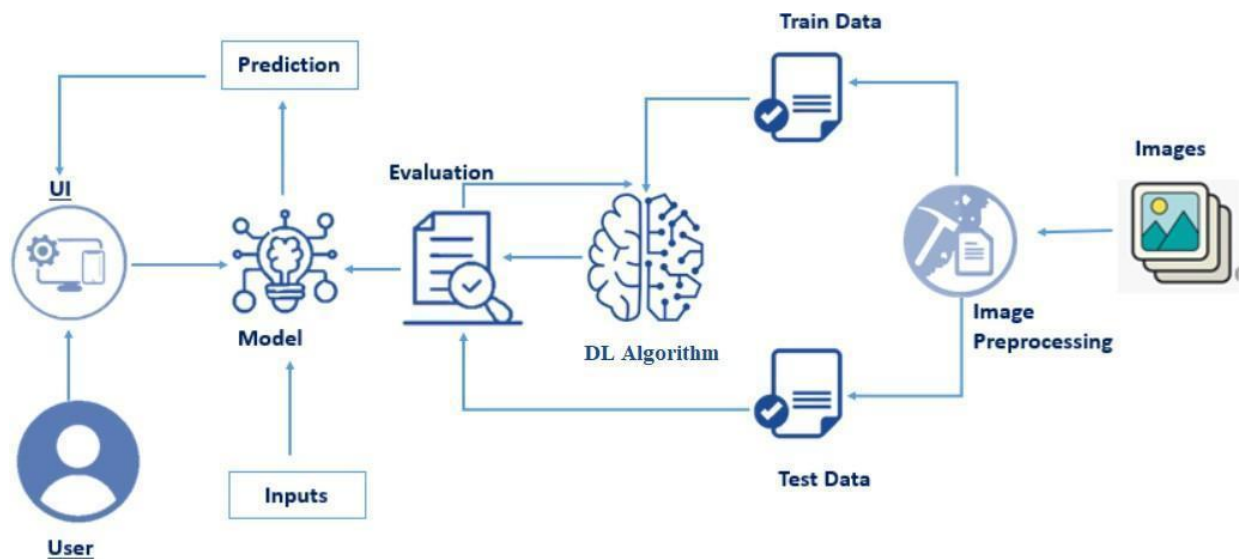
The dataset consists of images categorized into five classes. For each class, 600 images were selected. Images were resized to 224x224 pixels and normalized to [0,1]. Train-test-validation split was 70%-15%-15%. Data augmentation was avoided due to sufficient class balance.

## 1 . Model Architecture

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The model uses a MobileNetV2 feature extractor from TensorFlow Hub with a dense output layer of five units and softmax activation. It was compiled with Adam optimizer

and sparse categorical crossentropy loss. The model was trained for 10 epochs with validation monitoring.



## 1. Evaluation and Accuracy

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Evaluation was performed on a separate test set. Metrics included accuracy and classification report. The model achieved high classification performance with validation accuracy improving across epochs, as visualized via accuracy and loss plots.

## 1. Web Application

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A Flask web app was developed to serve the trained model. Users can upload rice grain images, and the app predicts the rice variety along with a confidence score. The app uses OpenCV for image preprocessing and TensorFlow for inference.

## 1. Results

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The application successfully identifies uploaded rice grains with clear predictions like 'Basmati' or 'Jasmine' and returns a confidence percentage. The system performs realtime image inference with high reliability and visual feedback through web UI.

## 1. Advantages & Limitations

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Advantages: Automated classification, user-friendly interface, scalable model deployment.

Limitations: Model trained on fixed image sizes and may not generalize to unseen lighting or image quality conditions.

## 1. Conclusion

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The rice classification project demonstrates the application of transfer learning in agriculture. The system can help automate grain classification and can be scaled with more data and advanced features.

## 1. Future Scope

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Future improvements include integrating image augmentation, mobile deployment, adding more rice varieties, and improving generalization through diverse datasets.