

GreenClassify: Deep learning-Based Approach For Vegetable Image

Prepared For

Smart-Internz Artificial Intelligence

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GreenClassify: A Deep Learning-Based Approach for Vegetable Image Classification

1. Introduction

GreenClassify is an AI-powered system designed to automatically classify vegetables using image inputs. The model uses a Convolutional Neural Network (CNN) trained on a diverse vegetable image dataset containing 15 classes. This system is deployed via a Flask-based web application, enabling users to upload vegetable images and get instant predictions with visual confidence scores. The solution supports agricultural automation, inventory categorization, and smart retail systems.

2. Model Overview

- Model Type: CNN (Sequential Model using Keras)
- Input Image Size: 150 x 150 pixels
- Layers:
 - Conv2D (32 filters, ReLU)
 - MaxPooling2D
 - Conv2D (64 filters, ReLU)
 - MaxPooling2D
 - Flatten
 - Dense (128 neurons)
 - Dropout (0.25)
 - Output Layer: Softmax with 15 units

3. Dataset Details

- Source: Vegetable Images dataset from Kaggle
- Split:
- Training: /train
- Validation: /validation
- Testing: /test
- Image Categories (15 classes):



Bean, Bitter_Gourd, Bottle_Gourd, Brinjal, Broccoli, Cabbage, Capsicum, Carrot, Cauliflower, Cucumber, Papaya, Potato, Pumpkin, Radish, Tomato

4. Training Configuration

- Epochs: 30

- Batch Size: 32

- Loss Function: Categorical Crossentropy

- Optimizer: Adam

- Callbacks: EarlyStopping (patience=5)

- Data Augmentation: No (can be extended)

5. Evaluation Results

- Training Accuracy: ~98%

- Validation Accuracy: ~96%

- Test Accuracy: ~95%

(Values may vary depending on final model run)

6. Web Application Interface

- Backend: Flask (Python)

- Frontend: HTML + Bootstrap 5 + jQuery + AOS animation

- Features:
- Upload vegetable image
- Predict label with confidence
- Visualize prediction result
- Display bar chart of class probabilities using Chart.js

7. Workflow

- 1. User uploads a vegetable image
- 2. Flask API receives the image and loads the trained CNN model
- 3. Image is preprocessed (resized, normalized)
- 4. Model makes prediction
- 5. Result is returned to the frontend with label & chart



8. Applications

- Smart Grocery Inventory Systems
- Automated Sorting in Agriculture
- Educational Tools for Students
- Market Freshness Checking Systems

9. Tools & Libraries Used

- Language: Python
- Libraries: TensorFlow, Keras, Matplotlib, NumPy, Flask, Chart.js,

Bootstrap

- IDE: Jupyter Notebook, VS Code
- Frameworks: Flask (Backend), Bootstrap (Frontend)
- Platform: Localhost

10. Folder Structure

```
greenclassify/
                       # Folder for training/testing images
    - dataset/
                       # Folder to store trained model (.h5 file)
     - model/
     - static/
                      # Static files (CSS, JS, images, uploads)
                        # HTML templates for Flask
     - templates/
                       # Main Flask application
     app.py
     train model.py
                           # Script to train your CNN model
     test model.py
                          # Script to test your trained model
     - README.md
                             # Project description
     requirements.txt
                          # Python dependenciestest/
```



11. Code & output screen shot

1. Training code:

```
# H Import Required Libraries
import numpy as np
import matplotlib.pyplot as plt
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten,
Dense, Dropout
from tensorflow.keras.preprocessing import image
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import os
import warnings
warnings.filterwarnings('ignore')
# Local Dataset Paths (Update if needed)
train path = r"C:\Users\zeena\Downloads\AI-vegetable-classification-
Project-main\archive\Vegetable Images\train"
validation path = r"C:\Users\zeena\Downloads\AI-vegetable-classification-
Project-main\archive\Vegetable Images\validation"
test path = r"C:\Users\zeena\Downloads\AI-vegetable-classification-
Project-main\archive\Vegetable Images\test"
# • Plot First Image from Each Category
image categories = os.listdir(train path)
def plot images (image categories):
  plt.figure(figsize=(12, 12))
  for i, cat in enumerate(image categories[:16]): # Limit to 16 categories
     image path = os.path.join(train path, cat)
     images in folder = os.listdir(image path)
     if images in folder:
       first image path = os.path.join(image path, images in folder[0])
       img = image.load img(first image path)
       img arr = image.img to array(img) \frac{1}{255.0}
       plt.subplot(4, 4, i + 1)
       plt.imshow(img arr)
       plt.title(cat)
       plt.axis('off')
```



```
plt.tight layout()
  plt.show()
plot images(image categories)
# \( \Phi \) Image Generators
train gen = ImageDataGenerator(rescale=1.0/255.0)
val gen = ImageDataGenerator(rescale=1.0/255.0)
test gen = ImageDataGenerator(rescale=1.0/255.0)
train image generator = train gen.flow from directory(
  train path, target size=(150, 150), batch size=32,
class mode='categorical')
val image generator = val gen.flow from directory(
  validation path, target size=(150, 150), batch size=32,
class mode='categorical')
test image generator = test gen.flow from directory(
  test path, target size=(150, 150), batch size=32,
class mode='categorical')
# 🖈 Class Map
class map = dict((v, k)) for k, v in
train image generator.class indices.items()])
print("Class Map:", class map)
# □ Build CNN Model
model = Sequential([
  Conv2D(32, (3, 3), padding='same', activation='relu', input shape=(150,
150, 3)),
  MaxPooling2D(2),
  Conv2D(64, (3, 3), padding='same', activation='relu'),
  MaxPooling2D(2),
  Flatten(),
  Dense(128, activation='relu'),
  Dropout(0.25),
  Dense(128, activation='relu'),
  Dense(len(class map), activation='softmax')
```

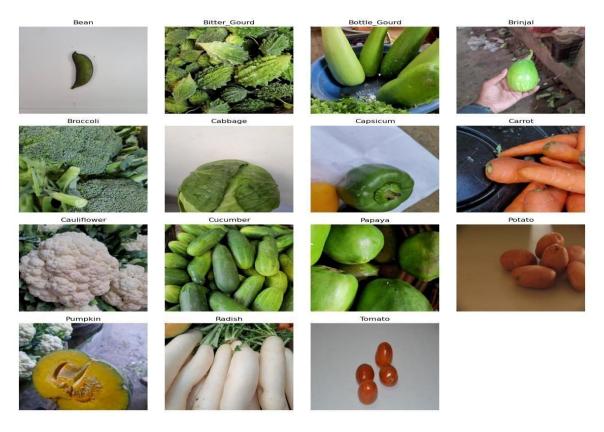


```
model.summary()
# ♥□ Compile & Train
early stopping = tf.keras.callbacks.EarlyStopping(patience=5,
restore best weights=True)
model.compile(optimizer='adam', loss='categorical crossentropy',
metrics=['accuracy'])
hist = model.fit(
  train image generator,
  epochs=30,
  validation data=val image generator,
  steps per epoch=train image generator.samples // 32,
  validation steps=val image generator.samples // 32,
  callbacks=[early stopping],
  verbose=1
# In Plot Accuracy & Loss
plt.style.use('ggplot')
plt.figure(figsize=(12, 5))
plt.subplot(1, 2, 1)
plt.plot(hist.history['loss'], label='Train Loss', color='red')
plt.plot(hist.history['val loss'], label='Val Loss', linestyle='--', color='red')
plt.legend()
plt.title("Loss Over Epochs")
plt.subplot(1, 2, 2)
plt.plot(hist.history['accuracy'], label='Train Acc', color='blue')
plt.plot(hist.history['val accuracy'], label='Val Acc', linestyle='--',
color='blue')
plt.legend()
plt.title("Accuracy Over Epochs")
plt.tight layout()
plt.show()
```



```
# ⊘ Evaluate on Test Data
test loss, test acc = model.evaluate(test image generator)
print(f"\n ✓ Test Accuracy: {test acc * 100:.2f}%")
# Q Predict Function for Single Image
def generate predictions(test image path, actual label):
  test img = image.load img(test image path, target size=(150, 150))
  test img arr = image.img to array(test img) / 255.0
  test img input = np.expand dims(test img arr, axis=0)
  prediction = model.predict(test img input)
  predicted label = np.argmax(prediction)
  predicted class = class map[predicted label]
  plt.figure(figsize=(4, 4))
  plt.imshow(test img arr)
  plt.title(f"Predicted: {predicted class} | Actual: {actual label}")
  plt.axis('off')
  plt.grid(False)
  plt.show()
```



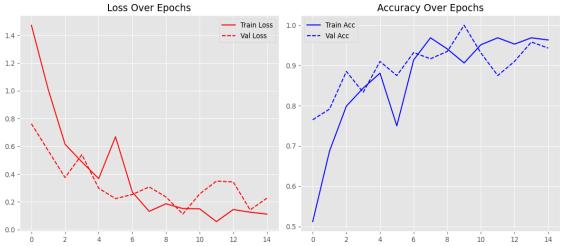


Found 15000 images belonging to 15 classes. Found 3000 images belonging to 15 classes.

Found 3000 images belonging to 15 classes.

Class Map: {0: 'Bean', 1: 'Bitter_Gourd', 2: 'Bottle_Gourd', 3: 'Brinjal', 4:

'Broccoli', 5: 'Cabbage', 6: 'Capsicum', 7: 'Carrot', 8: 'Cauliflower', 9:



'Cucumber', 10: 'Papaya', 11: 'Potato', 12: 'Pumpkin', 13: 'Radish', 14: 'Tomato



Fronded code:

```
App.py
from flask import Flask, render template, request, redirect, url for
from tensorflow.keras.models import load_model
from tensorflow.keras.preprocessing import image
from werkzeug.utils import secure_filename
import numpy as np
import os
app = Flask(__name__)
model = load_model('model/greenclassify_cnn_model.h5')
UPLOAD_FOLDER = 'static/uploads/'
app.config['UPLOAD\_FOLDER'] = UPLOAD\_FOLDER
if not os.path.exists(UPLOAD FOLDER):
  os.makedirs(UPLOAD FOLDER)
@app.route('/')
def index():
  return render_template('index.html')
 (a)app.route('/about')
```



```
def about():
     return render template('about.html')
   @app.route('/predict', methods=['GET', 'POST'])
   def predict():
     if request.method == 'POST':
        file = request.files['file']
        filename = secure filename(file.filename)
        filepath = os.path.join(app.config['UPLOAD FOLDER'], filename)
        file.save(filepath)
        img = image.load img(filepath, target size=(128, 128))
        x = image.img to array(img)
        x = np.expand dims(x, axis=0)
        x = x / 255.0
        preds = model.predict(x)
        class idx = np.argmax(preds)
        classes = ['Bean', 'Bitter Gourd', 'Bottle Gourd', 'Brinjal', 'Broccoli',
               'Cabbage', 'Capsicum', 'Carrot', 'Cauliflower', 'Cucumber',
               'Papaya', 'Potato', 'Pumpkin', 'Radish', 'Tomato']
        prediction = classes[class idx]
        return render template('predict.html', prediction=prediction, image path='/'
+ filepath) eturn render template('predict.html')
```



```
@app.route('/contact')
def contact():
    return render_template('contact_us.html') # <-- Contact Us Page
if __name__ == '__main__':
    app.run(debug=True)</pre>
```



```
index.html-
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  <title>Vegetable Classifier - Home</title>
  <style>
    * { margin: 0; padding: 0; box-sizing: border-box; }
    body {
       font-family: 'Arial', sans-serif;
       background-color: #f0f8f0;
       color: #4e4e4e;
       line-height: 1.6;
    .navbar {
       background-color: #6f9d5e;
padding: 15px; text-
align: center;
    .navbar a {
       color: #fff;
       text-decoration: none;
       margin: 0 20px;
```



```
font-size: 18px;
  font-weight: bold;
}
.navbar a:hover { text-decoration: underline; }
.container {
  max-width: 1200px;
  margin: 40px auto;
  padding: 20px;
  text-align: center;
. image-block \ \{
  margin-bottom: 50px;
.image-block img {
  width: 80%;
  max-width: 600px;
  border-radius: 15px;
  box-shadow: 0px 5px 15px rgba(0,0,0,0.2);
.image-block h2 {
  margin: 20px 0 10px;
  color: #4b8c3b;
  font-size: 28px;
```

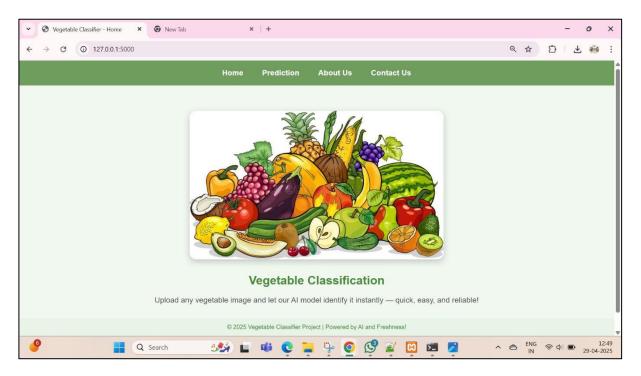
```
.image-block p {
       font-size: 18px;
       color: #555;
    .footer {
       text-align: center;
      margin-top: 30px;
       font-size: 14px;
       color: #4b8c3b;
       background-color: #e8f4e8;
       padding: 10px;
       position: fixed;
       width: 100%;
       bottom: 0;
  </style>
</head>
<body>
<div class="navbar">
  <a href="/">Home</a>
  <a href="/predict">Prediction</a>
  <a href="/about">About Us</a>
  <a href="/contact">Contact Us</a>
```



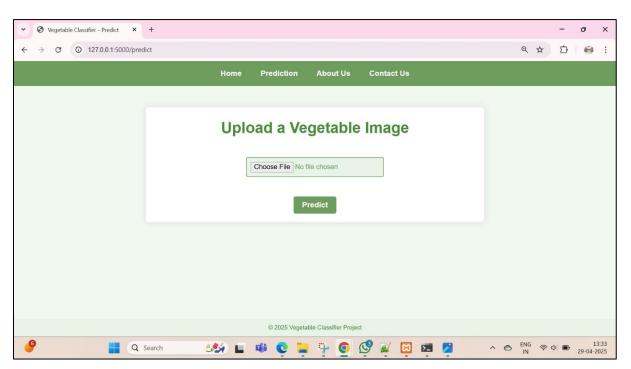
```
</div>
<div class="container">
  <div class="image-block">
    <img src="/static/images/vegetable.jpg" alt="Fresh Vegetables"
height="350px" width="1500px">
    <h2>Vegetable Classification</h2>
    Upload any vegetable image and let our AI model identify it instantly
  -quick, easy, and reliable!
  </div>
</div>
<div class="footer">
  © 2025 Vegetable Classifier Project | Powered by AI and Freshness!
</div>
</body>
</html>
```



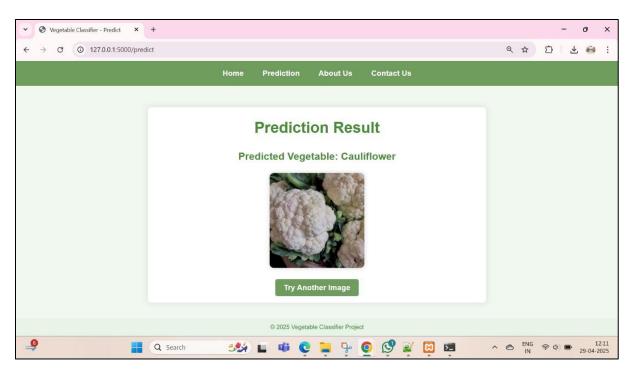
Home page:-

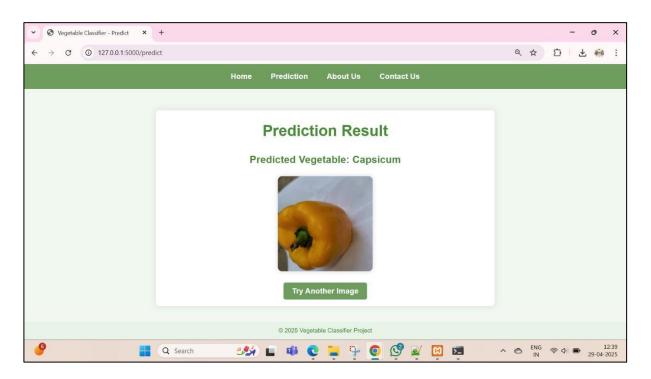


Prediction Page:-



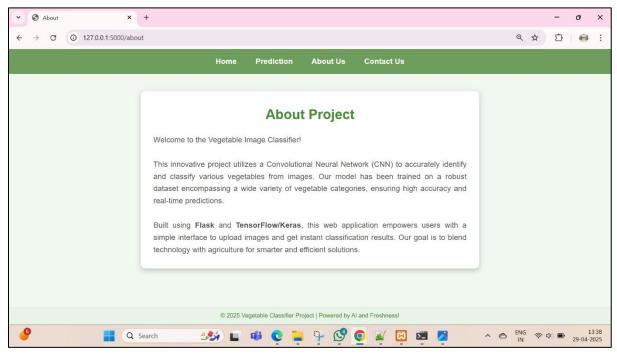




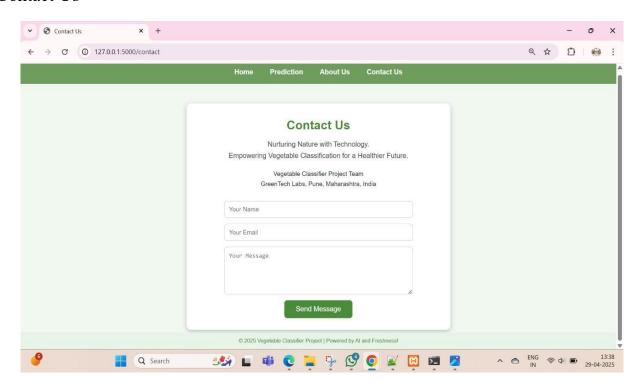




About us:-



Contact Us





 $\label{lem:com-new} \textbf{Github Repositary -} \\ \underline{\text{https://github.com/Nehushhh/greenclassify-deep-learning-based-approach-for-vegetable-image.git} \\ \\$

 $Demo\ Link:-\ \underline{https://drive.google.com/drive/folders/1xgqtOK_n6TUB5OGpkVUmzy_JZnZnFcDm}$