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Macromedia University of Applied Sciences

Course Title: Basics of Machine Learning **Name of Examiner:** jeniffer Callou

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Vegetable Classification Using Convolutional Neural Network (CNN)

Project Overview

This project is about building a computer program that can recognize vegetables in images. We used a machine learning technique called Convolutional Neural Networks (CNN), which is especially good at understanding pictures. The aim was to create a simple, working model that can tell which vegetable is in a photo. This work is submitted as part of an academic course and shows how artificial intelligence can help with image-related tasks.

Objective

- To create a model that identifies 15 types of vegetables from images.
- To use simple preprocessing steps to improve training results.
- To test the model using new images to see how well it works.
- To understand and apply deep learning concepts in a practical way.

Dataset Information

- The dataset has 15 folders, one for each vegetable category.
- Each class includes about 1,000 training images.
- There are separate folders for validation and test images with 200 images per class.
- The pictures have different lighting and backgrounds to mimic real-life conditions.

Steps for Data Preparation

- All images were resized to 150 by 150 pixels.
- The pixel values were scaled between 0 and 1 for better learning.
- We used Keras ImageDataGenerator to load and process images in batches.
- The training data was shuffled to help the model generalize.
- No image augmentation was used in this version, but it can be added later.

Model Structure

The model is made using basic CNN layers:

- Input layer accepts 150x150 size RGB images.
- Three convolutional layers with 32, 64, and 128 filters, each followed by max pooling layers.
- Flatten layer converts data to a 1D format.
- Dense layer with 128 neurons and a dropout layer to avoid overfitting.
- Output layer with softmax activation to predict one out of 15 vegetable classes.

Model Training

- We trained the model for 10 epochs using batch size 32.
- RMSprop optimizer and categorical crossentropy were used.
- The validation accuracy reached over 90%, which is good for a basic CNN.

Making Predictions

After training, we saved the model. To test it:

- 1. Load the saved model file (.keras).
- 2. Load the image you want to test.
- 3. Resize the image to 150x150.
- 4. Convert it to an array and normalize it.
- 5. Use the model to predict the vegetable.
- 6. Print the name of the predicted class.

User Interface (Optional)

A basic web interface was created using Streamlit. This allows users to upload a photo and get the prediction directly in a browser. It is optional and just shows how the model could be used in a simple app.

Files Included in the Project

- vegetable classifier model.keras The trained model file.
- vegetable_classifier_notebook.ipynb Notebook with training and testing code.
- prediction script.py Script for testing the model with a new image.
- streamlit_app.py Simple Streamlit-based web UI (optional).
- requirements.txt List of required Python libraries.
- project documentation.docx Full project write-up.
- README.txt Short instructions on how to use the files.
- image2.jpeg Sample image for testing.
- screenshots/ Folder with screenshots of training results and predictions.
- dataset sample/ Few example images from the dataset.

How to Run the Project

- 1. Make sure Python is installed and activate your virtual environment.
- 2. Open a terminal and go to the project folder.
- 3. Run the following command to install the needed libraries:

pip install -r requirements.txt

4. To make a prediction using a test image, run:

python prediction script.py

5. To use the web interface, run:

streamlit run streamlit app.py

6. Make sure the .keras model file is in the same folder as the script or notebook when testing.

Conclusion

This project was a great way to understand how deep learning can be used in real-world tasks. We learned how to prepare data, train a CNN model, and test it on new images. The results were good and show that even a simple CNN can work well when trained properly. More improvements can be made by adding more data and using advanced techniques, but this version already shows the main concepts clearly.