Project Plan: Object Recognition System

Introduction:

The object recognition project aims to develop a system capable of identifying objects in real-time using machine learning algorithms. This system will enable the identification of various objects captured by a laptop's camera.

Objective:

The primary objective of this project is to train a machine learning model to recognise different objects and classify them into predefined categories accurately.

Methodology:

1. Data Collection and Model Training:

- A dataset of images containing various objects is collected and used to train the machine learning model.
- The model is trained to recognise unique features and patterns associated with each object category.

2. Real-time Object Identification:

- The system uses the laptop's camera to capture images of objects in real-time.
- Captured images are processed using the trained model to identify the object category.

Implementation:

- The system is implemented using Python programming language and machine learning libraries such as TensorFlow and scikit-learn.
- The user interface allows users to interact with the system by uploading images for training and using the laptop's camera for real-time object identification.

Results:

- The trained model demonstrates high accuracy in identifying objects from various categories.
- Real-time object identification using the laptop's camera provides efficient and reliable results.

3. Modules to be Implemented:

- 1. Data Collection and Preprocessing
- 2. Model Trainina
- 3. Model Evaluation
- 4. Deployment
- 5. Object Recognition Testing

4. Week-wise Module Implementation and High-level Requirements

Week 1-2: Data Collection and Preprocessing

- Collect a dataset of images containing various objects for training and testing.
- Preprocess the dataset (e.g., resizing, normalisation).
- Output: Screenshots of the collected dataset and preprocessing steps.

Week 3-4: Model Training

- Select a machine learning algorithm (e.g., CNN) and train the model on the dataset.
- Output: Screenshots of the model training process and evaluation metrics.

Week 5: Model Evaluation

- Evaluate the trained model using a separate test dataset.
- Output: Screenshots of the model evaluation results and accuracy metrics.

Week 6: Deployment

- Deploy the trained model to a cloud or local environment for inference.
- Output: Screenshots of the deployed model in action, recognising objects in new images or video streams.

Week 7-8: Object Recognition Testing

- Design a user-friendly interface for uploading images or streaming video.
- Implement object recognition functionality using the deployed model.
- Test the system with a variety of images or video streams containing different objects.
- Output: Screenshots of the testing interface, recognised objects, and classification results.

5. Evaluation Criteria Milestone-wise:

- Milestone 1 (Week 4): Successful training of the model with a satisfactory accuracy rate.
- Milestone 2 (Week 5): Evaluation of the model with a high accuracy rate (>80%).
- Milestone 3 (Week 6): Deployment of the model with successful object recognition.
- Milestone 4 (Week 8): Testing of the object recognition system with accurate object identification and classification.

6. Design Diagram:

• Provide a high-level architectural diagram showing the components of the object recognition system, including data flow and processing steps.

- upload_image(image_data: Image) - display_result(result: Result) Storage and Deployment save_model(model: Model) load_model()

UserInterface

Data Processing Pipeline - upload_image(image_data: Image - preprocess_image(image: Image) - train_model() - optimize_model() - recognise_objects (image: Image) Output - display_objects(result)

Components:

1. User Interface (UI):

- Responsible for interacting with the user.
- Allows users to upload images or stream video for object recognition.
- Displays the recognition results to the user.

2. Data Processing Pipeline:

- **Preprocessing Module:** Receives images from the UI and preprocesses them (e.g., resizing, normalisation) to prepare them for object recognition.
- Model Training and Optimisation Module: Trains the machine learning model using the preprocessed images and optimises the model for better performance.
- **Object Recognition Module:** Receives preprocessed images from the preprocessing module and uses the trained model to recognise objects in the images.

3. Storage and Model Deployment:

- **Storage Module:** Manages the storage of the trained machine learning model and any necessary data.
- Model Deployment Module: Deploys the trained model to a suitable environment (e.g., cloud, local server) for inference during object recognition.

4. Output Module:

• Receives the recognition results from the object recognition module and displays them to the user through the UI.

Data Flow:

- 1. The user uploads an image or streams video through the UI.
- 2. The UI sends the image/video stream to the Data Processing Pipeline.
- 3. The Preprocessing Module preprocesses the image/video stream.
- 4. The Model Training and Optimisation Module trains and optimises the machine learning model.
- 5. The Object Recognition Module uses the trained model to recognise objects in the preprocessed image/video stream.
- 6. The recognition results are sent to the Output Module.
- 7. The Output Module displays the recognition results to the user through the UI.

Technology Stack:

- Language: Python
- Libraries/Frameworks: TensorFlow, Keras (for machine learning), Flask/Django (for web application), SQLAlchemy (for database)

7. Conclusion:

The completion of this project will result in a fully functional object recognition system, demonstrating the interns' understanding and practical application of machine learning algorithms in object classification.