Project Title

Binary Image Classification Using CNNs:

Distinguishing Between Two Object Categories (Cars vs. Bikes)

Introduction

With the rise of image-based AI applications, binary classification tasks play a pivotal role in applications like security surveillance, quality control, and autonomous vehicles. This project aims to build a Convolutional Neural Network (CNN) to classify images into two categories — cars and bikes — leveraging state-of-the-art machine learning techniques.

Objective

The primary objective of this project is to design, train, validate, and test a robust CNN model for accurate binary image classification. The proposed solution will achieve the following:

- Accurately classify images as either "car" or "bike."
- Optimize performance using hyperparameter tuning.
- Evaluate generalization through rigorous testing.

Scope

The project will encompass:

1. Dataset Preparation:

- Use a well-structured dataset containing images of cars and bikes, organized into train, validation, and test splits.
- Apply preprocessing and transformations for data augmentation.

2. Model Development:

• Build a CNN model using PyTorch, convolutional, pooling, and fully connected layers.

3. Training and Validation:

Train the model on the training set, with validation to monitor overfitting.

4. Testing:

• Evaluate the model on unseen test data.

5. Output:

• Save and visualize results with clear insights into model performance.

Technical Approach

1. Technology Stack:

Framework: PyTorch

• Libraries: TorchVision, NumPy, Matplotlib, Tqdm, Streamlit

• Hardware: GPU for accelerated training (if available)

2. Model Architecture:

- Input: Images resized to 224x224.
- Layers:
 - Convolutional layers with ReLU activation.
 - Max-pooling layers for dimensionality reduction.
 - Fully connected layers for classification.
- Output: Single node

3. Training Strategy:

- BCEWithLogitsLoss.
- Adam optimizer for gradient updates.
- Batch size: 32
- Learning rate: 0.001
- Epochs: 5 (adjustable based on results)

Expected Outcomes

- **Functional CNN Model**: Capable of classifying images with high accuracy.
- **Metrics**: Achieve at least 90% accuracy on the test set.
- Scalability: The model should be adaptable for multi-class classification tasks.

Resources Required

- 1. **Hardware**: A machine with GPU support for accelerated training (NVIDIA CUDA preferred).
- 2. Software:
 - Python 3.8+
 - Required Python libraries (torch, torchvision, numpy, matplotlib, tqdm).