Temporal Hand Gesture Recognition for Renesas RA8D1

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Built For Renesas RA8D1 Memory <1MB SRAM Accuracy 98%+
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1. Project Overview

This project delivers a real-time temporal hand gesture recognition system optimized for the Renesas RA8D1 microcontroller. It features a custom-built Temporal Convolutional Network (TCN) designed to operate within a strict 1MB SRAM memory constraint, demonstrating the successful deployment of advanced machine learning on a resource-constrained embedded platform.

The system is now fully functional and robust. After a rigorous debugging process that resolved critical bugs in the C-based Adam optimizer and the server's model-loading lifecycle, the entire pipeline—from data collection to training and real-time inference—operates with high accuracy and stability.

2. Getting Started

Prerequisites

- A C compiler (e.g., gcc or clang)
- make
- Python 3.11 (This specific version is required for GUI dependencies on macOS).

Automated Execution

The entire application is managed by a single master script. This command builds the C code, sets up the Python environment, starts the C server, and launches the GUI in the correct order.

```
# Clone the repository and run the application
git clone https://github.com/WillForEternity/PerClassLoRA--RA8D1.git
cd PerClassLoRA--RA8D1
./start_app.sh
```

The script handles all dependencies and ensures the C server is running before launching the GUI. It also gracefully shuts down all processes on exit.

Manual Execution (Advanced)

For developers who prefer manual control, the C backend can be compiled and run using the provided Makefile.

```
# Navigate to the C backend directory
cd RA8D1_Simulation

# Build both the training and inference executables
make all

# To run the executables (after building)
./train_c  # Run the training process
./ra8d1_sim  # Run the inference server

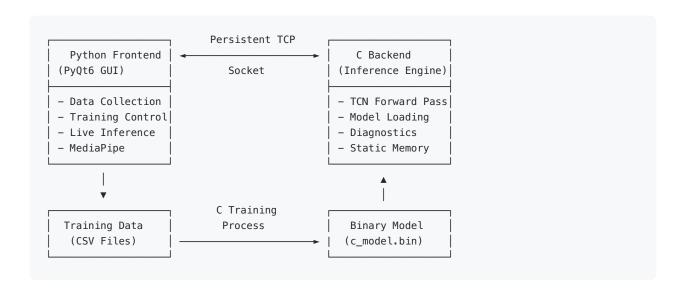
# To clean all build artifacts
make clean
```

3. Workflow

- 1. **// Initial Setup**: Run ./start_app.sh to automatically build all components and launch the GUI.
- Data Collection: Use the Data Collection tab to record temporal gestures using the live camera feed. Each gesture is captured as a 20-frame sequence and saved as a CSV file in models/data/{gesture_name}/.
- 3. Model Training: Navigate to the Training tab and click "Start Training." This invokes the train_c executable, which loads the CSV data, runs the training process for 500 epochs to ensure robust learning, and saves the final c_model.bin.
- 4. **©** Real-time Inference: Open the Inference tab to see live gesture recognition. The GUI sends normalized hand landmark data to the ra8d1_sim server and displays the returned prediction and confidence score.

4. System Architecture & Technical Specifications

The system employs a hybrid architecture, leveraging a Python-based GUI for user interaction and a high-performance C backend for model execution.



Components

Component	Technology	Role
GUI Frontend	Python (PyQt6)	Manages user workflow, data collection, and visualization.
Backend Engine	С	Handles all ML computation (training and inference) with static memory.
Communication	TCP Sockets	Persistent connection for low-latency binary data streaming.

Model & Data Pipeline

Aspect	Specification	
Model Type	Custom Temporal Convolutional Network (TCN) with 8 channels.	
Input Shape	20 frames × 63 features (21 landmarks × 3 coordinates).	
Activation	Leaky ReLU (α=0.01) to prevent dying neurons.	
Optimizer	Adam with He Initialization.	
Temporal Sampling	Stride-5 for both training and inference to ensure consistency.	
Memory Footprint	< 1MB SRAM (Static allocation, verified at compile-time).	
Performance	>98% accuracy; >30 FPS inference speed.	

5. Project Structure