

Phase 1: Technical Plan Presentation

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Background & Objectives

- Design, implement, and evaluate a neuralnetwork model for hand gesture recognition.
 - Compare model performance against established methods.
 - Document comprehensive methodology, datasets, and results.
 - ▶ Deliver final report and presentation with comparative analysis.

Algorithm Selection & Rationale



1. CNN + Temporal Pooling

- Simplifies temporal modeling by aggregating features without complex recurrence.
- Leverages frame-wise feature learning for clear spatial representation.



2. 3D-CNN (C3D)

- Learns joint spatiotemporal features directly from frame volumes.
- Provides a performance benchmark for more advanced models.



3. Transfer Learning+ LSTM (Optional)

- Utilizes pretrained visual representations to reduce training time.
- LSTM refines temporal dependencies for improved sequence modeling.

Dataset Selection & Rationale

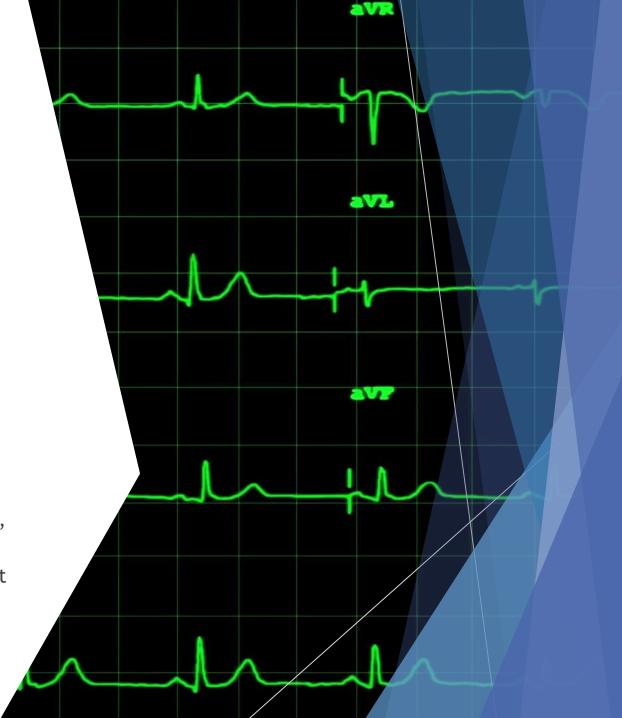
- Primary: LeapGestRecog
 - Provides balanced classes across 10 gestures ensuring fair evaluation.
 - Frame-based format eliminates video decoding complexity.
 - Moderate dataset size (~2,000 sequences) enables rapid iteration on standard hardware.
- Secondary (Future Phases):
 - SHREC dataset adds depth modality for richer feature exploration.
 - Custom webcam data tests model robustness to real-world variance.

Libraries & Technical Details

- Python 3.8+, TensorFlow2.x (Keras API)
 - OpenCV, NumPy & pandas for data handling
 - scikit-learn for splitting & metrics
 - Matplotlib & Seaborn for visualization
 - TensorBoard for monitoring
 - flake8 for code style compliance

Experimental Plan

- ▶ 1. Preprocess & split dataset (70/15/15 stratified)
- 2. Train CNN + Temporal Pooling baseline
- 3. Train C3D baseline for spatiotemporal comparison
- 4. Conduct hyperparameter sweep (LR, batch size, pooling method)
- ▶ 5. Evaluate on test set and compare metrics



Visualization & Metrics

- ► Learning curves: loss & accuracy over epochs
- Normalized confusion matrix heatmap
- Bar charts of per-class precision, recall, F1
- Sample frame sequences with predicted labels
- Efficiency table: parameters, FLOPs, inference time

Implementation & Reproducibility

Fixed random seeds for NumPy, TF, and Python random Configuration files for hyperparameters in configs/

Command-line scripts for data prep, training, evaluation

Version control via GitHub/GitLab repository