



BMS COLLEGE OF ENGINEERING

(Autonomous Institute, Affiliated to VTU, Belagavi)

DEPARTMENT OF MACHINE LEARNING

(UG Program: B.E. in Artificial Intelligence and Machine Learning)

Course : Video Analytics using Open CV

Course Code: 24AM6PEVCV

SIGN LANGUAGE FOR DEAF AND DUMB

Presented By,

Student Name & USN :

Archit Subudhi (1BM21AI026)

Aryaman Sharma (1BM21AI027)

Ayush Kumar Dubey (1BM21AI028)

Chetna Mundra (1BM21AI036)

Semester & Section: **6A**

Faculty In-Charge:

Dr. Seemanthini K

Assistant Professor

Department of Machine Learning

BMS College of Engineering

Introduction

- Importance of sign language in communication.
- Challenges faced by the deaf and hard-of-hearing community.
- Overview of sign language recognition systems.
- Advancements in computer vision and machine learning.



HELLO



GOODBYE



PLEASE



THANK YOU

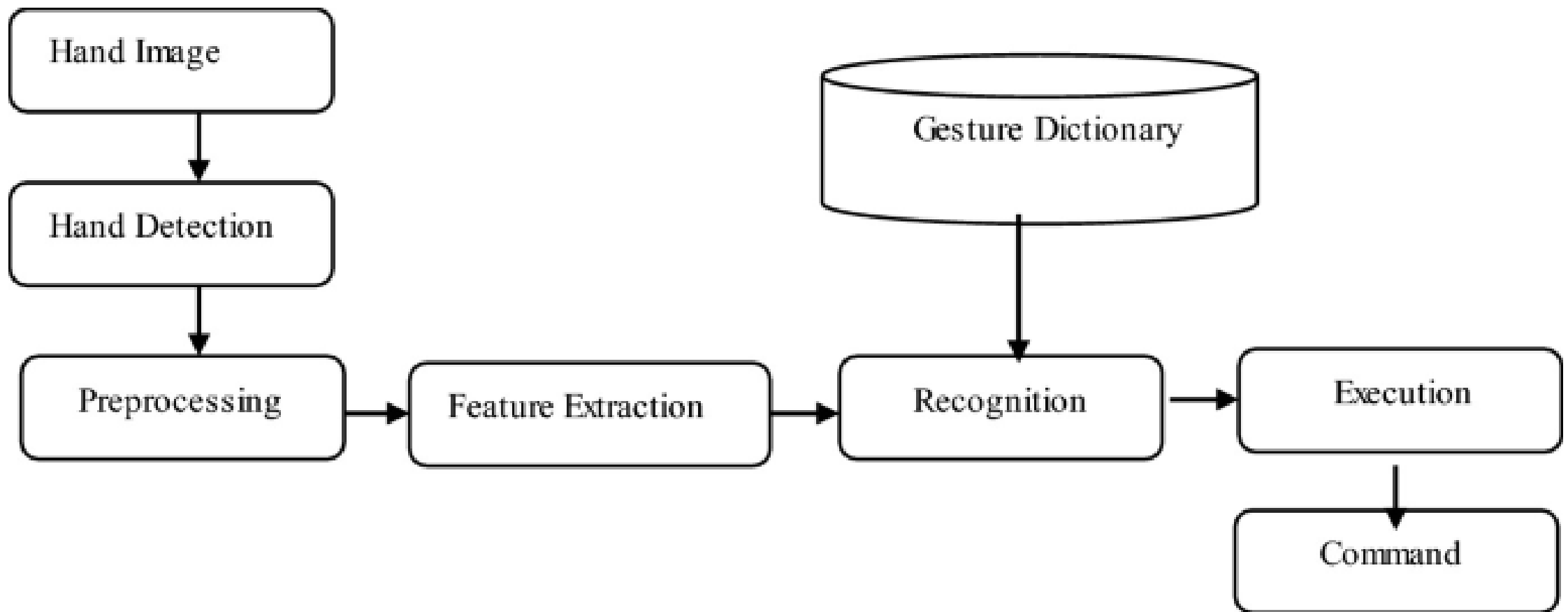


YES



NO

Design architecture



Mathematical Analysis

- Formulation of hand detection as a classification problem.

$$B = f_{\theta}(\text{image})$$

- Preprocessing steps: normalization and augmentation.

$$I' = P(I)$$

- Feature extraction: landmark identification.

$$L = \{(x_1, y_1, z_1), (x_2, y_2, z_2), \dots, (x_{21}, y_{21}, z_{21})\}$$

Mathematical Analysis

- Recognition and execution steps based on gesture dictionary.

$$g = M(L, G)$$

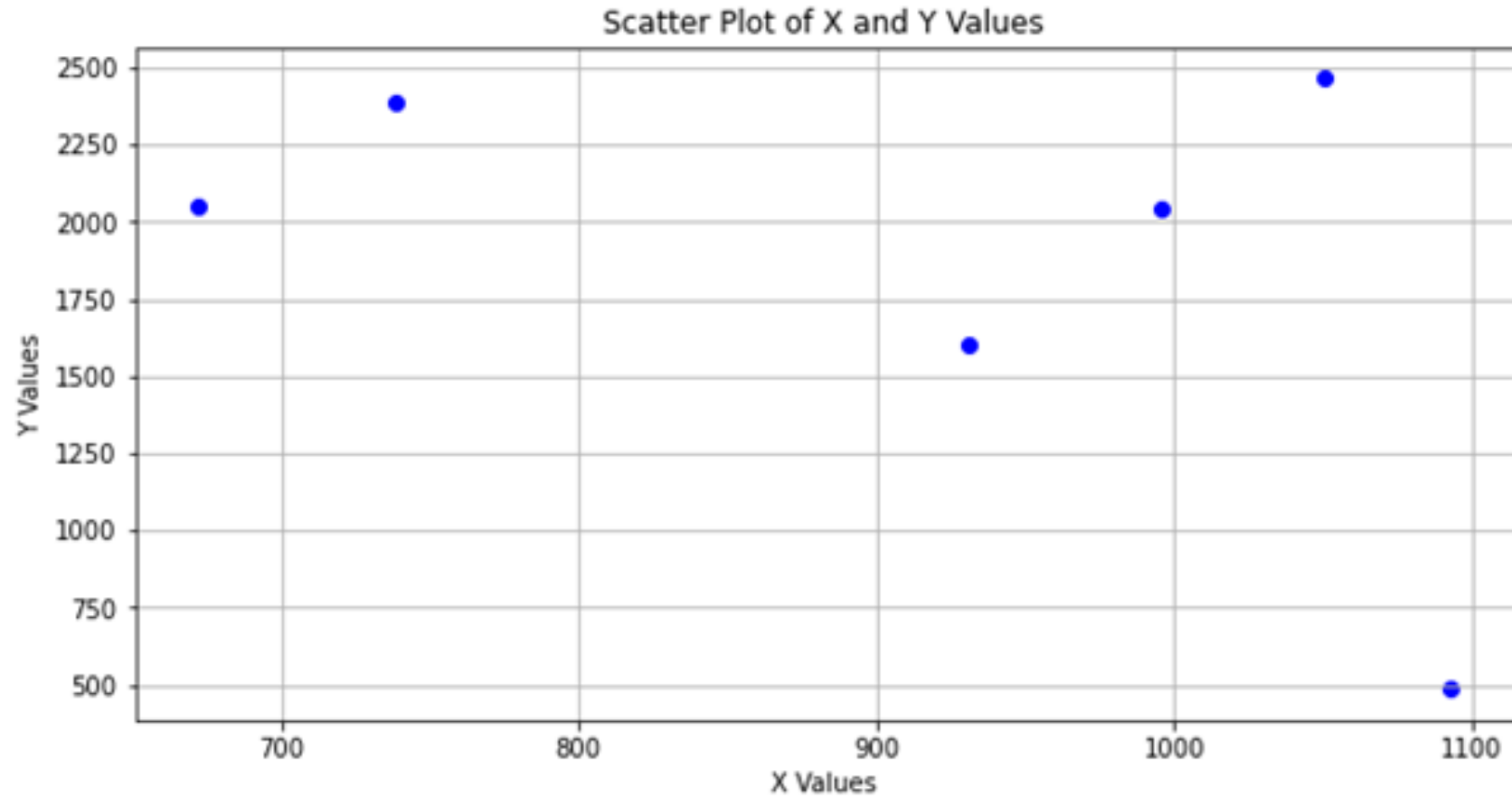
$$\mathcal{L}(\theta) = \sum_i \text{loss}(M(L_i; \theta), g_i)$$

$$C = \text{execute}(g)$$

Implementation

- Dataset creation using OpenCV and MediaPipe.
 - Tools and functions used: OpenCV (cv2), MediaPipe (mp).
- Details of data preprocessing: normalization and averaging.
 - Normalization and averaging of landmark positions.
 - Visualization using Matplotlib for normalized gestures.

Implementation



Implementation

- Training and testing: Euclidean distance calculation and nearest-neighbor classification.
 - Calculation of Euclidean distance between gestures.
 - Classification based on the closest match in the preprocessed dataset

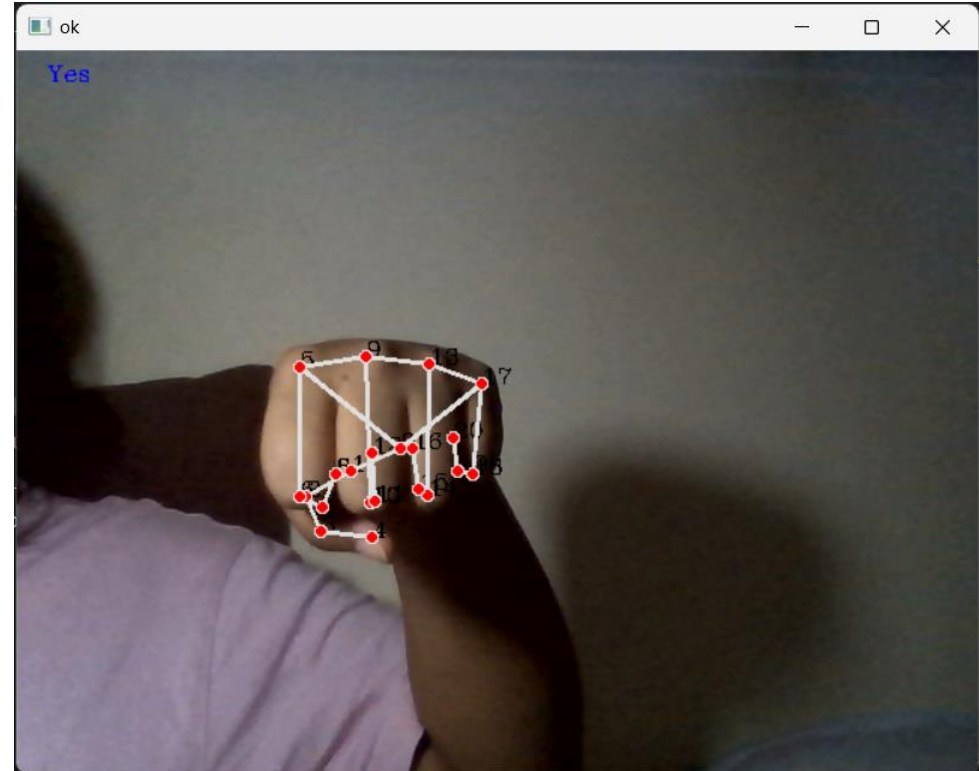
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1330.4449049595516  
679.28588336725  
814.178601245456  
goodbye  
[693, 357]  
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1710.2184838561657  
1267.6623820679947  
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109.40113828445979  
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223.54755503433103  
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goodbye
```


Implementation

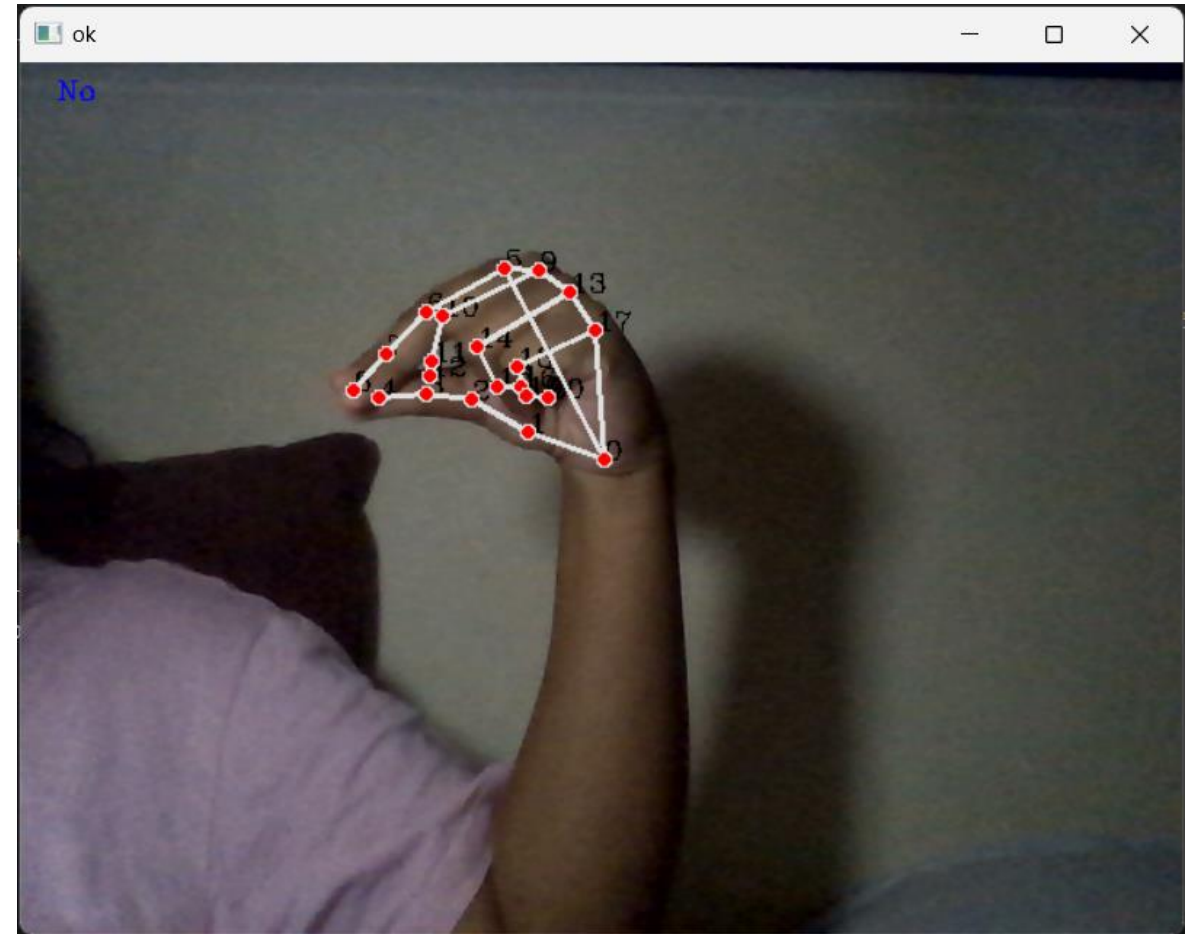
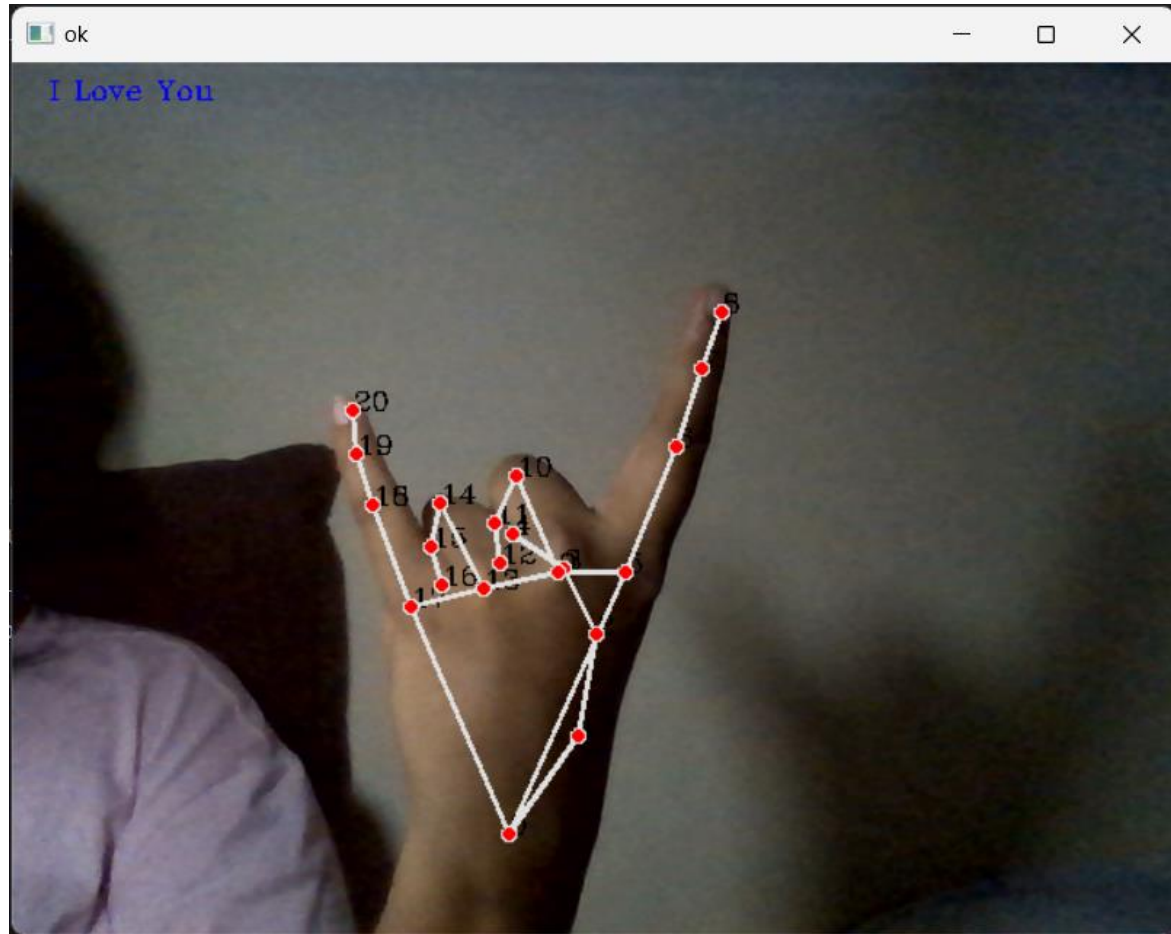
- Real-time prediction using MediaPipe and OpenCV.
 - Data Postprocessing
 - Integration of hand gesture recognition with nearest-neighbor approach.
 - Real-time display of recognized gestures on webcam feed.

Results and Analysis

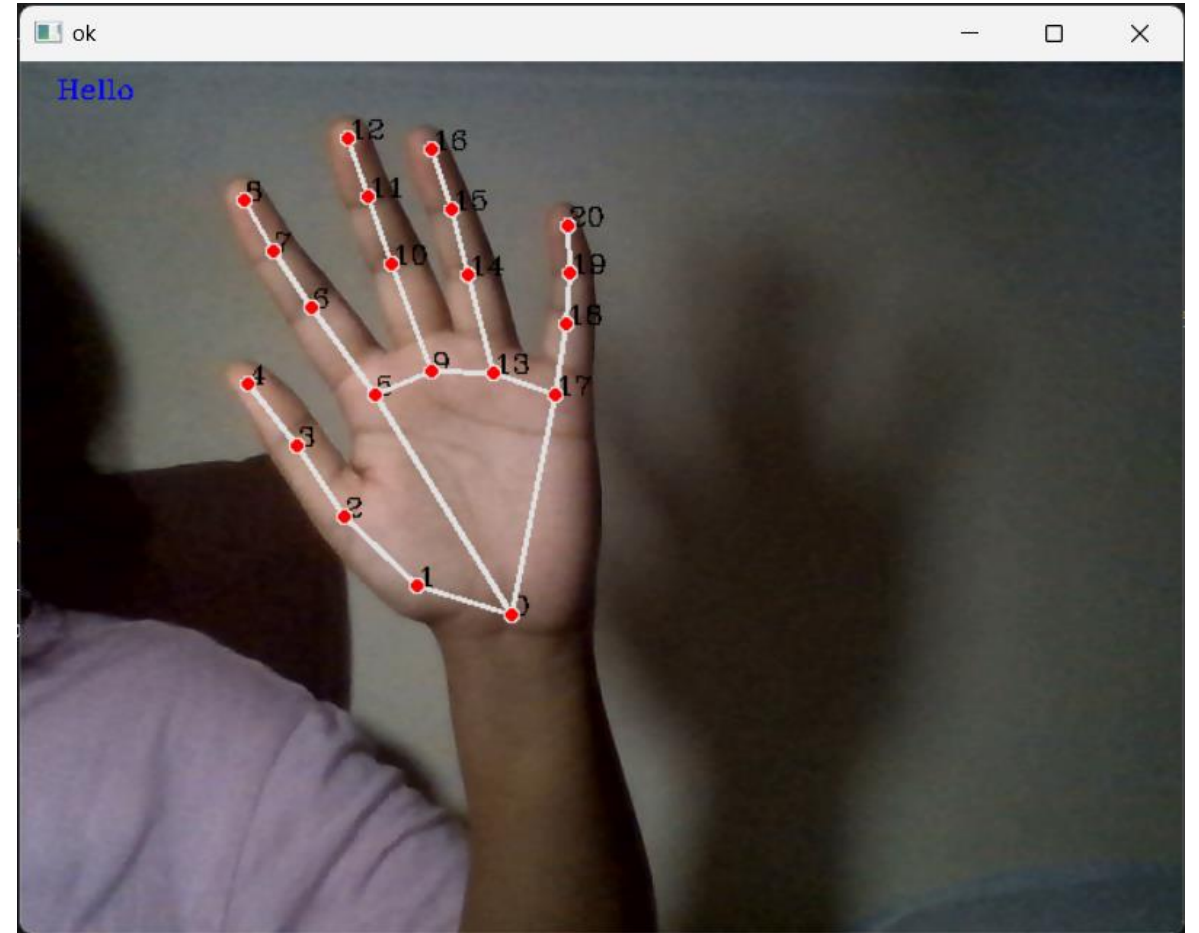
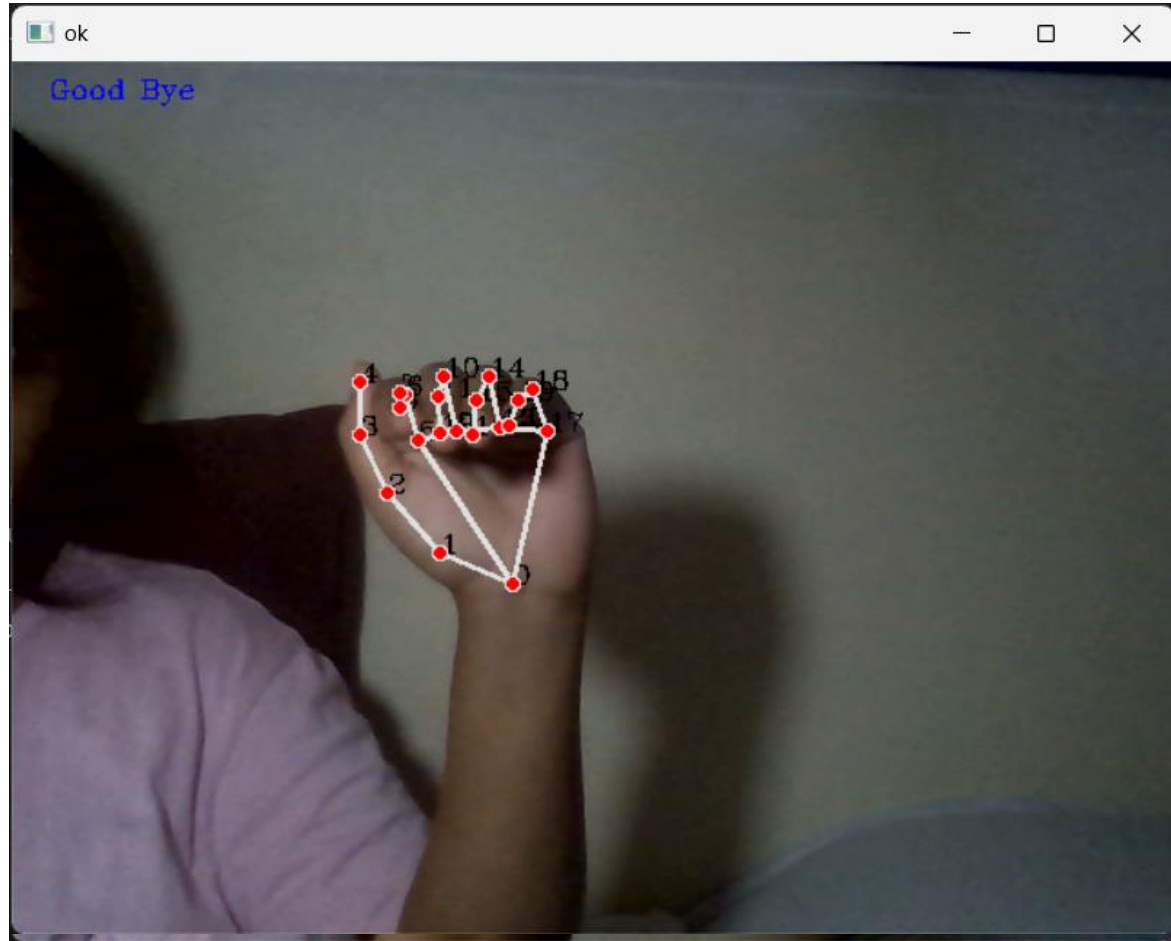
- Effectiveness of the real-time hand gesture recognition system.
- Accuracy of gesture classification using the nearest neighbor approach.
- Potential metrics for performance evaluation: accuracy, precision, and recall.
- Demonstration of system's robustness across various gestures.



Results and Analysis



Results and Analysis



Conclusion and Future Enhancement

- Impact on bridging communication barriers and enhancing accessibility.
- Potential for widespread adoption in various domains.
- Future enhancements: integration with deep learning models (CNNs, RNNs), dynamic gesture recognition, contextual information incorporation.
- Improving scalability and adaptability across different environments and user preferences.

Thank you !