# Hand Gestures Recognition Using 3D LiDAR Sensor

#### **Team Members:**

- Shruti Dilip Kavishwar,
- Xin Wen,
- Emily Weng,
- Jiani Ma,
- Feven Belay Araya

#### **Instructors**

- Prof. Henry Chang
- Prof. Alex Yang



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#### Introduction

#### Objective:

 Explore and evaluate diverse technologies for enhancing user interaction via gesture recognition.

#### **Approaches and Tools:**

 Implement GestureEVK and advanced CNN models with GUI integration for real-time application.

#### Goals:

- Evaluate the effectiveness and accuracy of each approach.
- Identify the best methods for robust and accurate gesture recognition.

### Three Approaches of Gestures Recognition

#### First Approach

Gesture Kit & Hand
Posture Al Model
with User Selection

#### **Second Approach**

3D Convolutional Neural Network Al Model

#### Third Approach

STMicroelectronic Merged Solution









#### **First Approach**

- Gesture Kit and Hand Posture Al Model with User Selection



#### **Dynamic Gestures:**

Palm Up Palm Left
Palm Down Palm Right

Tap Double



#### **Static Gestures:**

Palm Thumbs Up
Stop Thumbs Down

Heart Cross

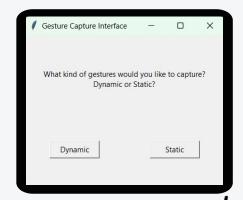
Fist

### First Approach – Basic Concept

- Use software GestureKit to recognize dynamic gestures and HandPosture Al Model to recognize static gestues
- Combined two softwares using User
   Selection Buttons Dynamic and Static,
   to switch between two softwares.

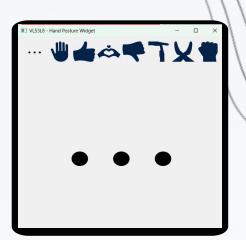


### First Approach – Process



**User Selection** 





Hand Posture Widget

**Dynamic Gestures Detection** 

### First Approach – Specification

**Distance:** The test subject is positioned at arm's length from the sensor. Hand gestures are performed 15 cm away from the sensor.

Duration: Each gesture demonstration lasts approximately 2 seconds.

**Interval:** There is a 3-second interval between demonstrations.

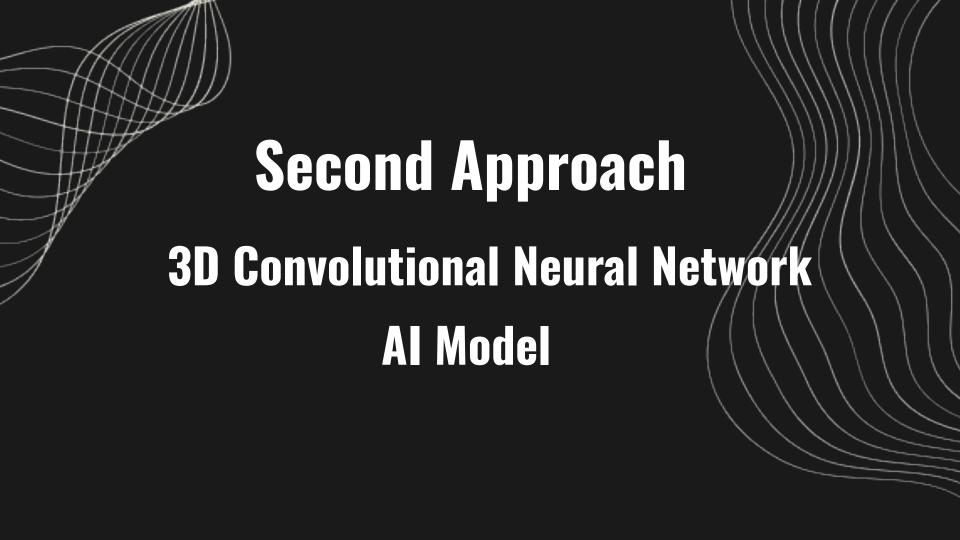
### **TEST ACCURACY**

<b>Dynamic Gestures</b>	****
Dynamic	
palm up	88.00%
palm down	99.00%
palm left	97.00%
palm right	96.00%
tap	95.00%
palm backward	92.00%

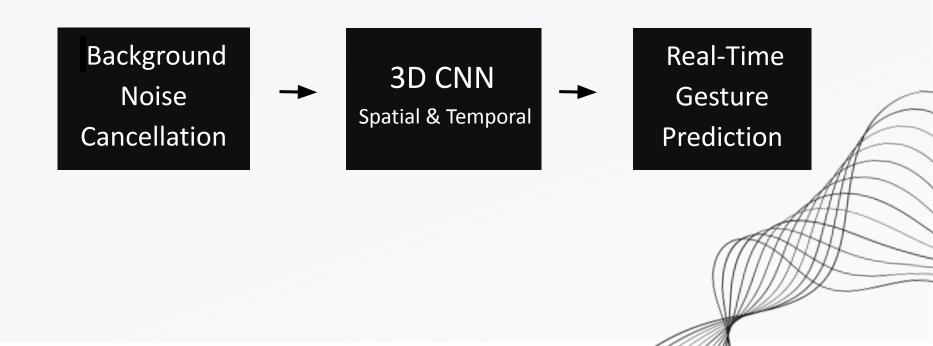
Static Gestures	****	
Static		
thumb up	98.00%	
tumb down	95.00%	
heart	93.00%	
stop	95.00%	
cross	99.00%	
fist	92.00%	







### **3D CNN AI Model - Basic Logic and Process**



### 3D CNN Al Model - Gesture Categories



**Dynamic Gestures:** 

Palm Up

Palm Left

Palm Forward

Palm Down

Palm Right

Palm Backward



**Static Gestures:** 

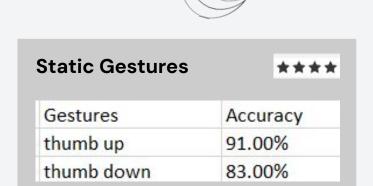
Thumbs Up Thumbs Down

### **Guidelines For Performing Gestures**

Gesture	Distance (cm)	Specifications
Palm Up	10~30	Move upward steadily. Centered entry and exit.
Palm Down	10~30	Move downward steadily. Centered entry and exit.
Palm Left	10~30	Move left steadily, no vertical movement.
Palm Right	10~30	Move right steadily, no vertical movement.
Palm Forward	10~30	Push forward steadily, no swaying.
Palm Backward	10~30	Pull backward steadily, no swaying.
Thumbs Up	25~35	Stationary, thumb up. Tilt hand slightly to sensor for clear heatmap. Left hand preferred.
Thumbs Down	25~35	Stationary, thumb down.

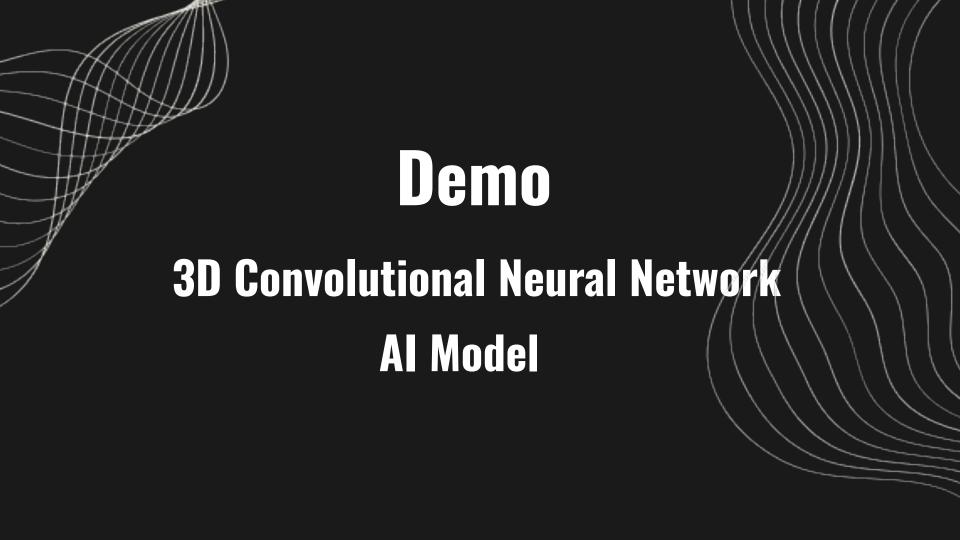
### **TEST ACCURACY**

Dynamic Gestures	****	
Gestures	Accuracy	
palm right	94.00%	
palm left	93.00%	
palm up	96.00%	
palm down	95.00%	
palm forward	98.00%	
palm down	98.00%	







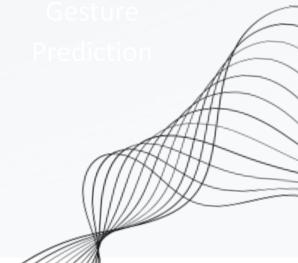




### STMicroelectronic Merged Solution

#### Limitation of First approach = Motivation for third approach

In reality we would want both the static and dynamic gestures to be recognized without any user intervention or user having to select which gesture they would want to perform. Having this in mind we have implemented a merged solution **using** the state of the art solution.



### STMicroelectronics Merged Solution

#### Logic

Sensors feeds data to Static and Dynamic models

```
If dynamic
gesture == True{
parse }
Else{ Parse static }
```

STATIC 0 == No Gesture / Idle If STATIC != 0 then parse

- Static Gestures are like images. Every frame in a video/feed will have a static gesture predicted.
- Dynamic Gesture is predicted only when the model detects change in motion
- So for this solution is a simplified solution using both static and dynamic models.

### **STMicroelectronic Merged Solution**

Dynamic Gesture Recognition Algorithm



Static Gesture
Al Model



Merged Solution

### **STMicrocontroller Merged Solution**



#### **Dynamic Gestures:**

\_eft

**Forward** 

Double Tap

**Down** 

Right

**Backward** 

#### **Static Gestures:**



**Flat Hand** 

Thumbs Up

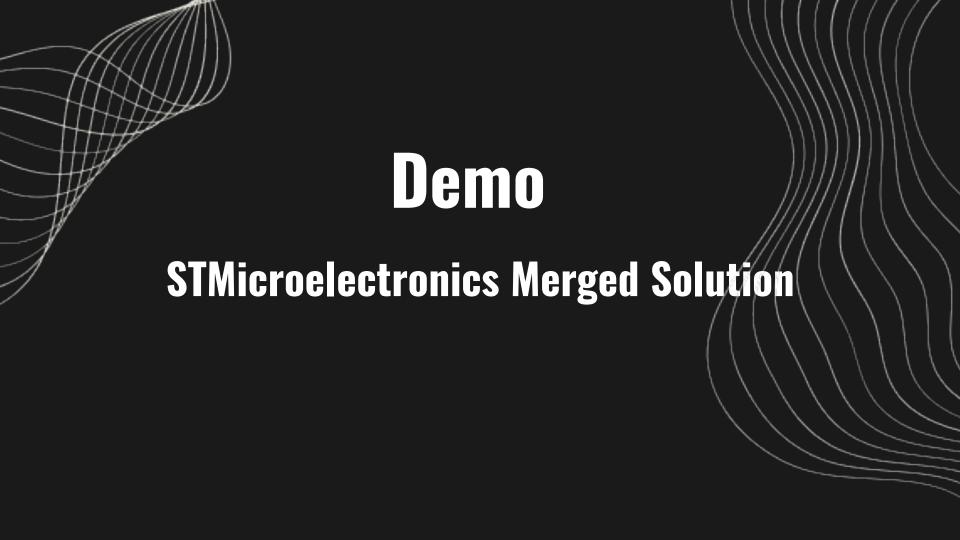
Heart

**BreakTime** 

**Thumbs Down** 

**Cross Hands** 

**Fist** 



### **Comparison of Test Accuracy**

Approach	Dynamic Gestures Accuracy (Avg)	Static Gestures Accuracy (Avg)	Key Strengths	Limitations
User Selection	~70%	~85%	Easy integration with GUI	Manual calibration needed
3D CNN Al Model	~95%	~87%	Advanced motion tracking high accuracy	Further improvements
STMicroelectronics Merged Solution	~80%	~90%	Advanced motion tracking	Higher computational complexity

# Fourth Approach

Prompt Based Gesture Recognition (using GPT API)

### **Prompt Based Gesture Recognition**

This project implements a gesture recognition system that uses real-time data from sensors to identify hand gestures. It combines:

- Sensor Data Collection
- Data Preprocessing and Analysis using pattern recognition methods
- Gesture Prediction with Prompt based solution
- Tkinter-Based GUI for Visualization

### **Prompt Based Gesture Recognition**

#### **Key Components**

- 1. Data Parsing and Frame Handling
  - a. Collects and parses raw sensor data into 8x8 frames.
- 2. Preprocessing
  - a. Extract meaningful features from collected frames
    - i. Mean, Std deviation, max, min distances
    - ii. Center of gravity
- 3. Real-Time Data Collection and Processing
  - a. Connects to serial port to collect and preprocess frames in real-time
- 4. Gesture Detection with Trends
  - a. Analyzes trends in CoG values for gesture prediction
  - b. Recognizes motion based on slopes
  - c. Identifies static and proximity-based gestures
- 5. GPT Integration

### **Prompt Based Gesture Recognition**

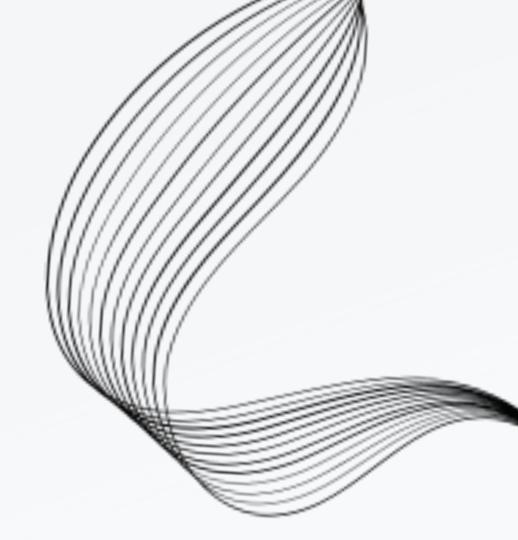
#### Highlights

- 1. Real-Time Gesture Analysis: Efficient frame collection and preprocessing pipeline
- 2. Trends and Al Refinement: Combines statistical analysis with GPT for high accuracy

#### **Example Gestures**

- 1. Upward motion: Detected when CoG moves significantly upward
- 2. Palm Forward: Identified based on proximity to the sensor.
- 3. Static Gesture: Small fluctuations with clear trends

## **THANKS**



### 2D&3D CNN AI Models - Basic Logic and Process

2D\_CNN
Spatial Only



Preprocess

3D\_CNN
Both Spatial
and Temporal



**Motion Detection** 

route input to appropriate model



**Evaluate** 

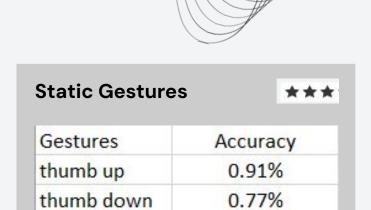


Static Gestures
Prediction

Dynamic Gestures Prediction

### **TEST ACCURACY**

Gestures	Accuracy
palm right	91.00%
palm left	83.00%
palm up	100.00%
palm down	91.00%
palm forward	71.00%
palm down	88.00%







#### Presentation slides:

1. Project of Gesture Recognition

By: name

- 2. 3 Approaches of Gesture Recognition:
  - a. STMicroelectronics' STSW-IMG035\_EVK and STM32CubeAl combined source code with  $\mbox{\rm GUI}$
  - b. STMicroelectronics' STSW-IMG035\_EVK and STM32CubeAl with Prompt
  - c. 3D CNN & 2D CNN Model

- 3. Combined source code with GUI
- 4. Real-time test accuracy

STMicroelectronics' STSW-IMG035 EVK and STM32CubeAI with Prompt