

## Outline



Background and Motivation



**Devices Used** 



Goals and Objectives



Tools and Workflow



Challenges and Insights

## Background and Motivation

#### Advantages of Using Radar for Action Recognition

- a. Unaffected by lighting conditions.
- b. Preserves user privacy, unlike cameras that capture direct images of people.
- c. Capable of penetrating certain materials, such as curtains or clothing.

### **Application Scenarios**

- a. Elderly care environments.
- b. Indoor human tracking.
- c. Posture monitoring (e.g., sitting posture detection).

## **Devices Used**



#### AWR1642 BOOST-ODS

Automotive mmWave radar system



AWR1642



## DCA1000 (data capture)

- Captures raw ADC data (I/Q signals)
- Transfers data to PC via Ethernet



DCA1000

# Goals and Objectives

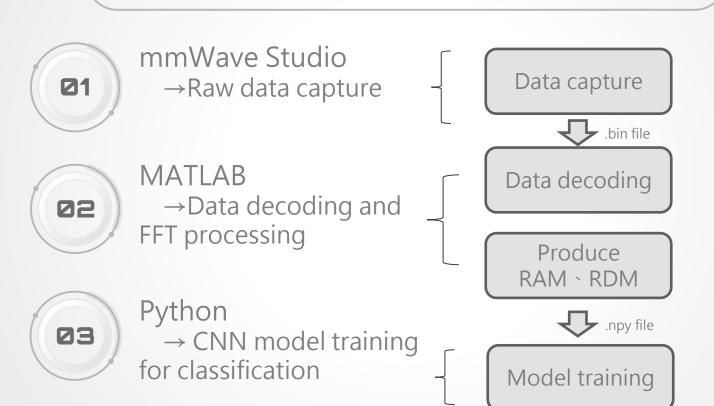
## Goal

 Use raw radar data to perform the full pipeline: from signal processing to model training.

## Setup

- 5 actions, 3 angles
  - Actions: standing, sitting, raising hands, hands on hips, using a smartphone.
  - Angles: -45°, 0°, +45°
- Single-person target
- Fixed environment

## Tools and Workflow



# mmwave Studio

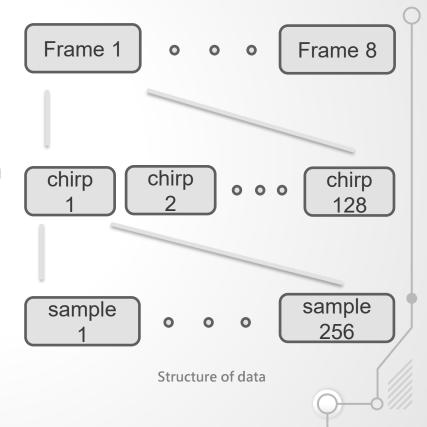
## mmwave Studio

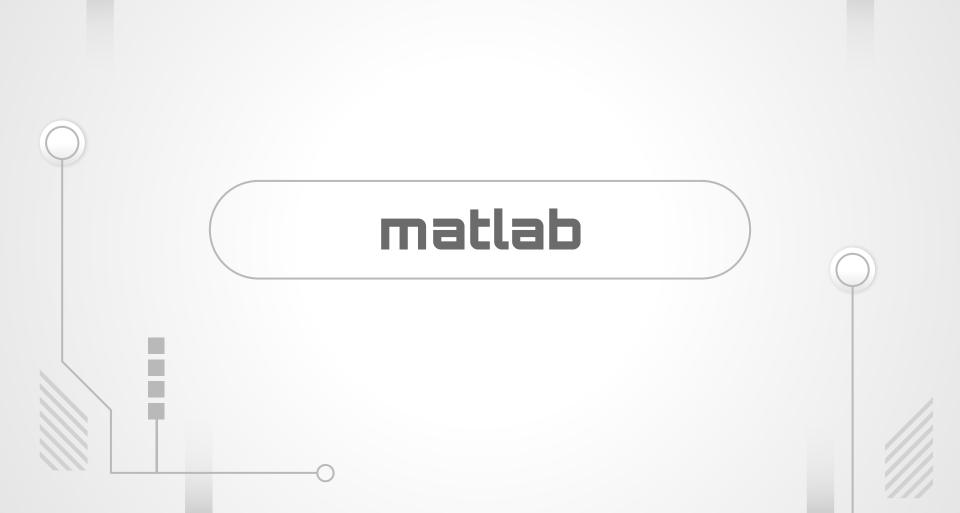
## Radar Configuration and Data Capture

- Set radar parameters: chirp and frame configurations
- Use DCA1000 to capture raw data through hardware interface
- Save data as .bin files (binary I/Q format)

## **Lua Script Automation**

 Lua scripts are used to automate radar parameter setup and trigger data capture





## matlab

#### a. Raw Data Reconstruction

 Parse .bin file into a 4D array: [Samples, Rx, Chirps, Frames].

#### b. 1D FFT (Range FFT)

- Perform FFT across ADC samples for each Rx and chirp.
- Converts time-domain signal into range information.
- Output: Range Profile (1D).

#### c. 2D FFT (Doppler FFT)

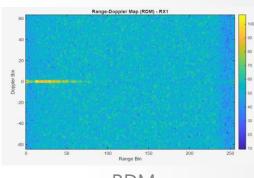
- Perform FFT across chirps (Doppler dimension)
- Output: Range-Doppler Map (RDM).

#### d. 3D FFT (Angle FFT)

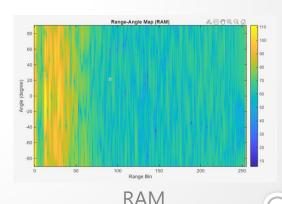
- Perform FFT across Rx channels (antenna array)
- Output: Range-Angle Map (RAM).

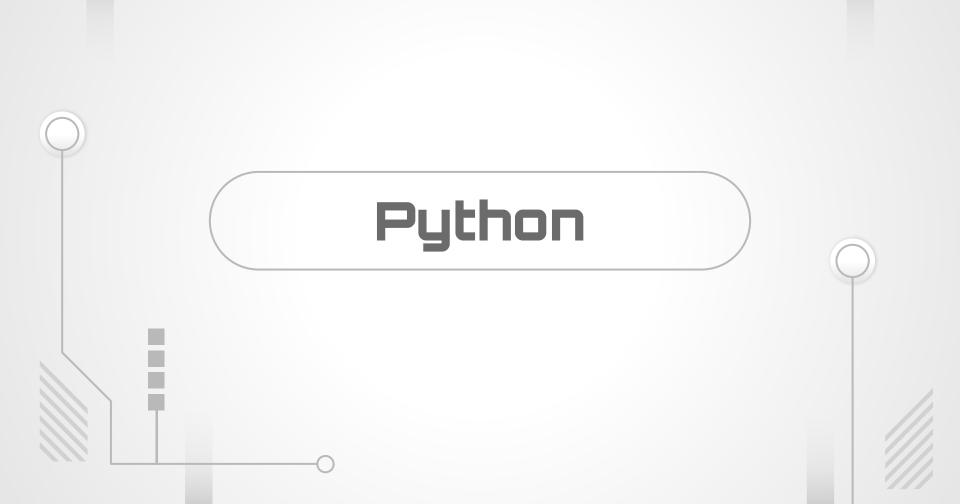
#### e. Data Storage

• Save processed results as .npy files for later use.



**RDM** 





# Python

#### **Input Data**

- Use .npy files of Range-Angle Maps (RAM) and Range-Doppler Maps (RDM)
- Explore a dual-input architecture

#### **Model Architecture**

- Built with TensorFlow using a Convolutional Neural Network (CNN)
- Structure: Conv → MaxPooling → Dense → Softmax

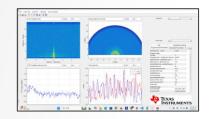
#### **Classification Targets**

- Action only
- or Action + Angle

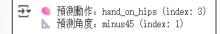
# Python











	accuracy	loss
僅分類動作	≈ 0.8	≈ 0.3

#### Issue

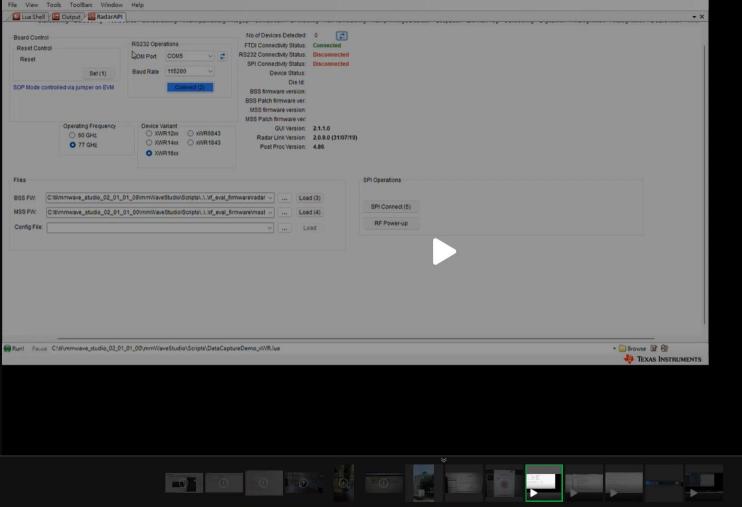
 The model struggles to recognize both action and angle simultaneously.

#### **Possible Reasons**

- Insufficient data volume.
- Too many classification categories.
- Data quality may be suboptimal.

#### **Potential Improvements**

- Increase dataset size.
- Adjust angle categorization strategy.
- Apply additional preprocessing (e.g., contrast enhancement).















































#### Challenges

- Collecting radar data.
- Operating and configuring mmWave radar hardware.
- Processing raw ADC data into meaningful features.
- Understanding and implementing the machine learning pipeline.

#### Insights

- Gained experience with the end-to-end workflow from raw radar data to model training.
- Developed understanding of the characteristics and applications of mmWave radar data.



