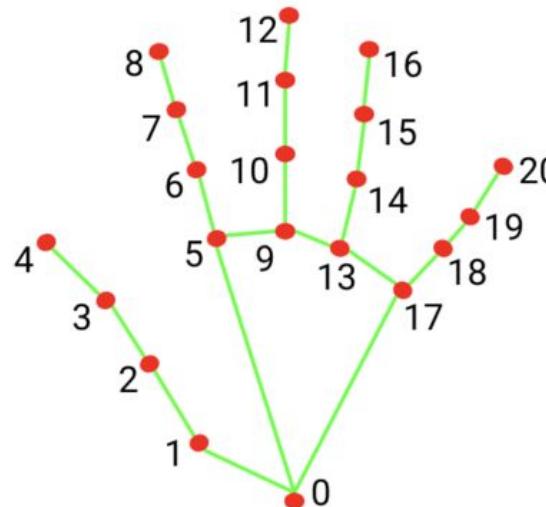
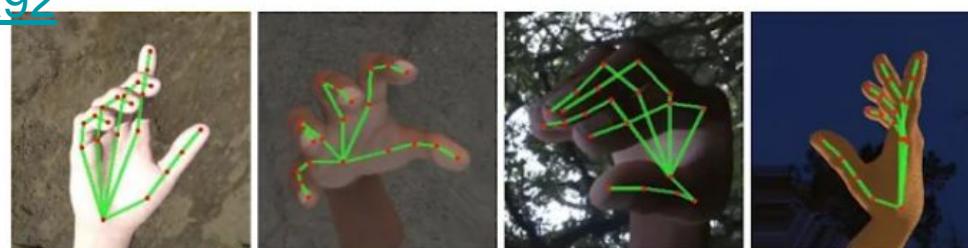


Light Controlling Using Hand Gestures



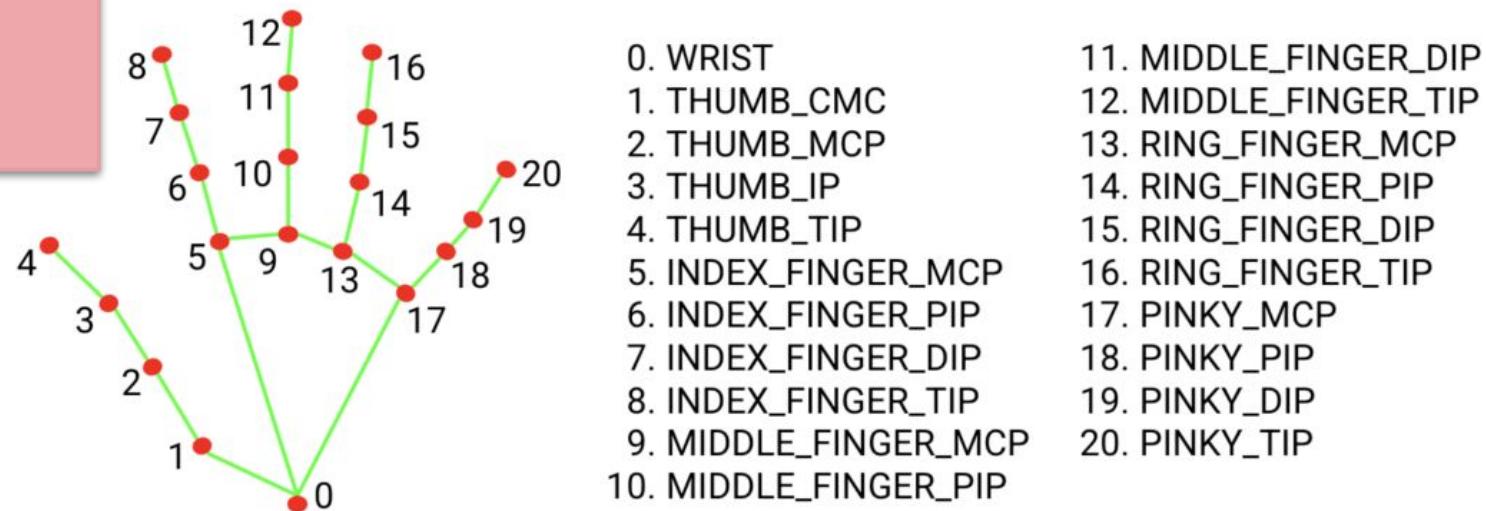
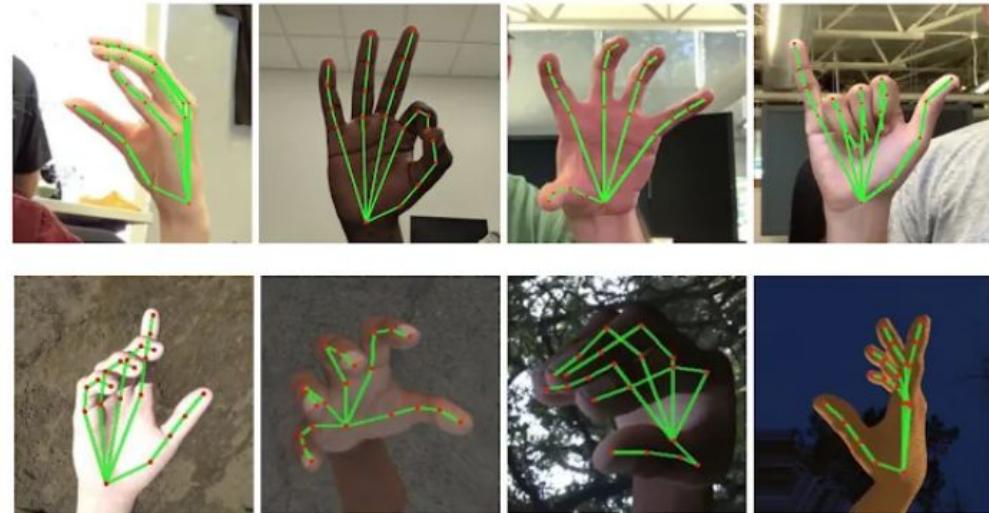
<https://chatgpt.com/c/67271026-a92c-8011-a73e-2a58da8f60ca>



- | | |
|-----------------------|-----------------------|
| 0. WRIST | 11. MIDDLE_FINGER_DIP |
| 1. THUMB_CMC | 12. MIDDLE_FINGER_TIP |
| 2. THUMB_MCP | 13. RING_FINGER_MCP |
| 3. THUMB_IP | 14. RING_FINGER_PIP |
| 4. THUMB_TIP | 15. RING_FINGER_DIP |
| 5. INDEX_FINGER_MCP | 16. RING_FINGER_TIP |
| 6. INDEX_FINGER_PIP | 17. PINKY_MCP |
| 7. INDEX_FINGER_DIP | 18. PINKY_PIP |
| 8. INDEX_FINGER_TIP | 19. PINKY_DIP |
| 9. MIDDLE_FINGER_MCP | 20. PINKY_TIP |
| 10. MIDDLE_FINGER_PIP | |

Objectives

- Develop a Natural Interaction System
- Set Up the Project Environment
- Collect and Prepare Gesture Data
- Build and Train an MLP Model
- Evaluate Model Performance
- Deploy Real-Time Gesture Recognition



Outline

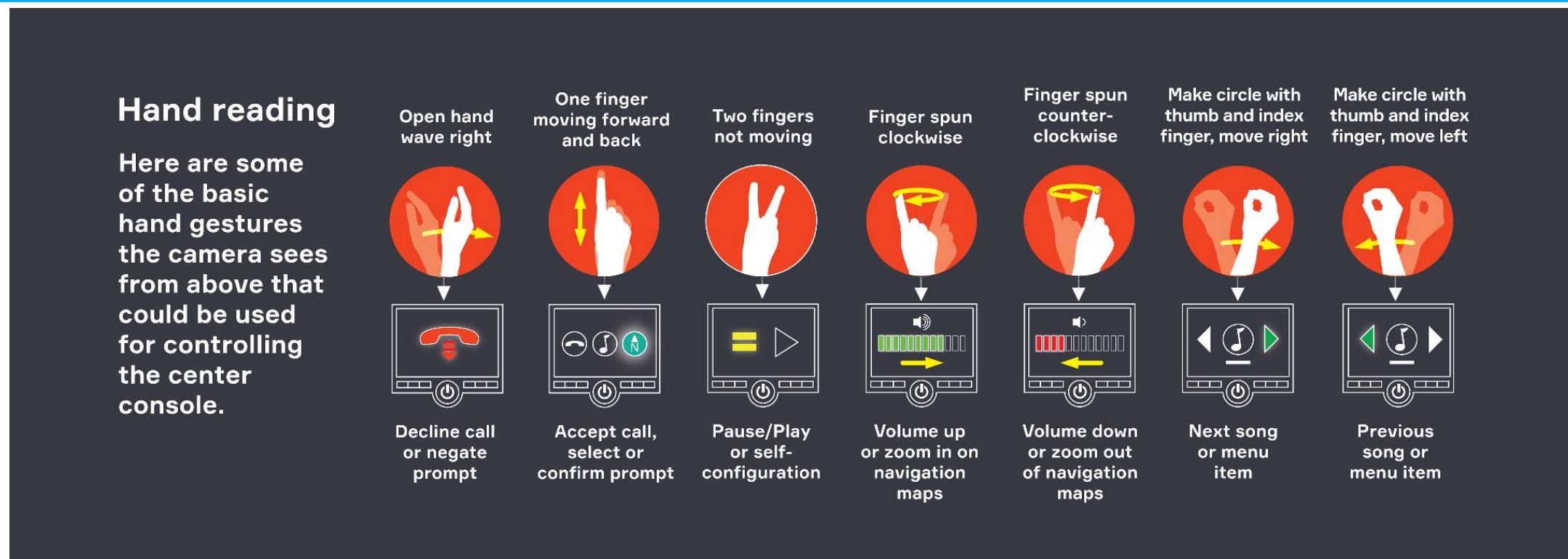
- 1. Hand Gestures Recognition**
- 2. Project Overview**
- 3. Step 0: Generate Data**
- 4. Step 1: Hand Gesture**
- 5. Step 2: Control Lights (Simulation + Reality)**
- 6. AI Application in Internet of Things (IoT)**

Hand Gestures Recognition

➤ What Is Hand Gestures Recognition?

Definition: HGR is technology that identifies and interprets hand and finger movements to enable intuitive interaction between humans and computers or robotic systems.

Significance: It enhances human-computer interaction (HCI) and human-robot interaction (HRI), with applications in healthcare, education, entertainment, and more.



Hand Gestures Recognition

➤ What Is Hand Gestures Recognition?

Key Components of HGR:

1. Data Acquisition:

- Utilizes cameras or sensors (standard video cameras or specialized depth cameras) to capture visual signals of hand gestures.

2. Feature Extraction:

- Processes captured images to identify key features such as hand positions, shapes, and movements using computer vision algorithms.

3. Classification:

- Applies machine learning or deep learning algorithms (e.g., MLP, CNN, SVM) to classify gestures into predefined categories.

Types of Hand Gestures:

- **Static Gestures:**

- Fixed hand positions like pointing or forming specific shapes.
- Recognition focuses on identifying these positions at a specific moment.

- **Dynamic Gestures:**

- Movements of hands over time, such as waving or drawing patterns.
- Recognition requires tracking the motion paths of the hands.

Hand Gestures Recognition

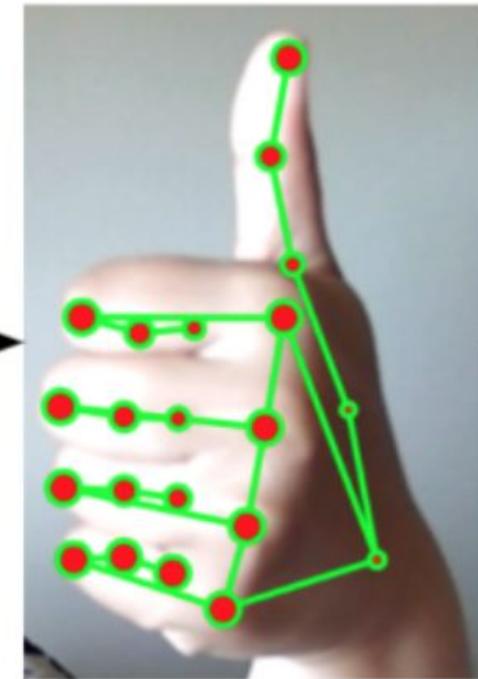
➤ What Is Hand Gestures Recognition?



RGB Image



Hand Skeleton



Gesture



Hình 1: Hand Gesture Recognition

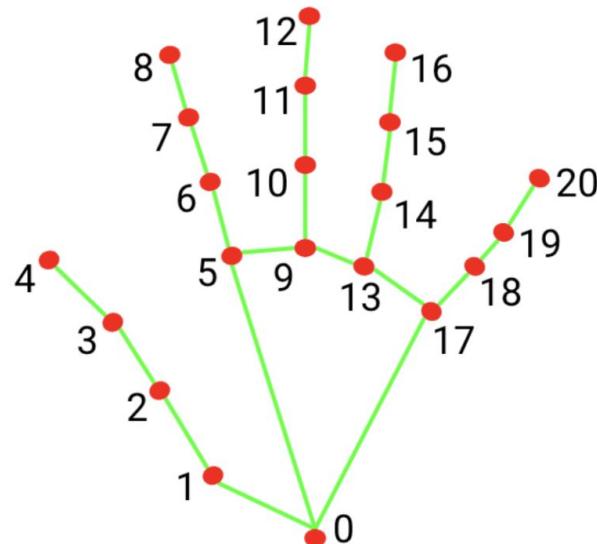
Hand Gestures Recognition

➤ What Is Hand Gestures Recognition?

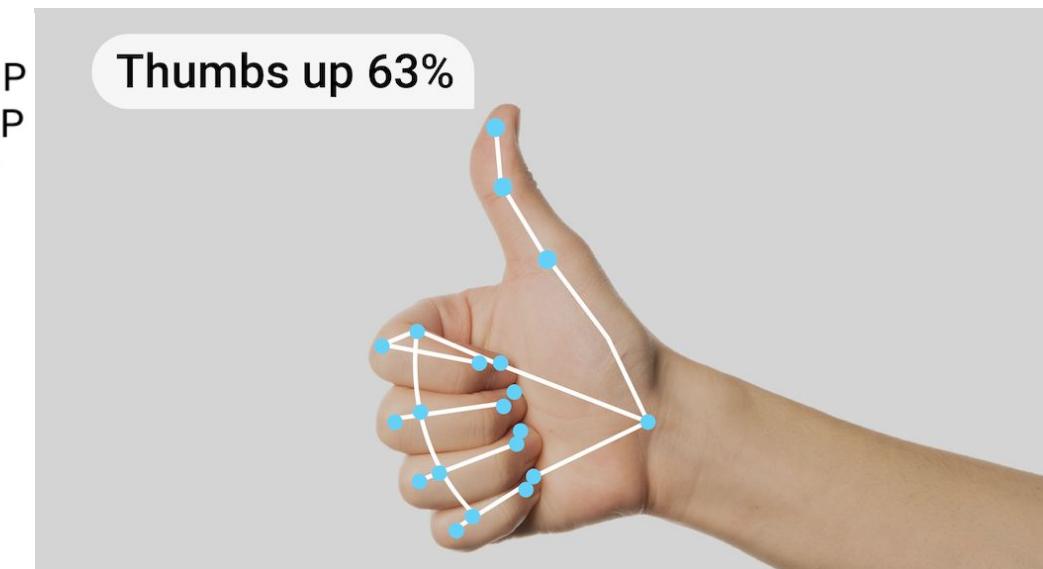
Mediapipe: palm detection model and hand landmarks detection mode

Hand landmark detection model:

- The keypoint localization of 21 hand-knuckle coordinates
- Model was trained on approximately 30K real-world images

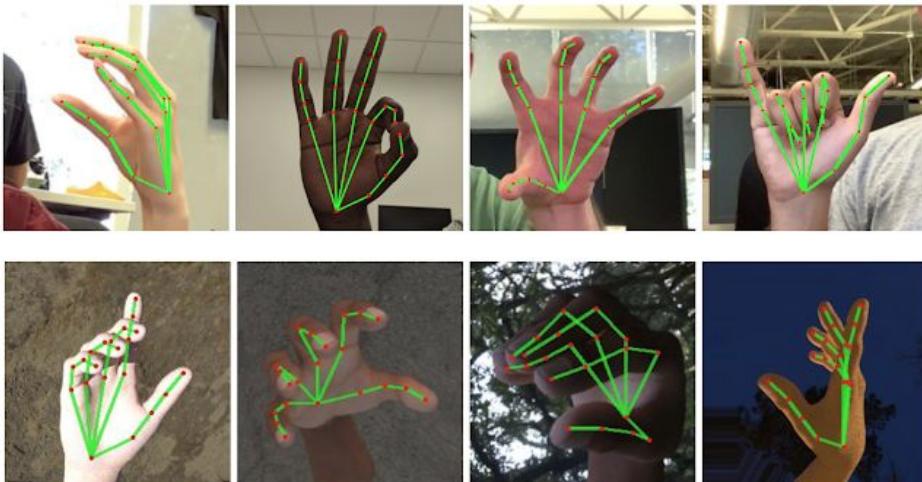


- | | |
|-----------------------|-----------------------|
| 0. WRIST | 11. MIDDLE_FINGER_DIP |
| 1. THUMB_CMC | 12. MIDDLE_FINGER_TIP |
| 2. THUMB_MCP | 13. RING_FINGER_MCP |
| 3. THUMB_IP | 14. RING_FINGER_PIP |
| 4. THUMB_TIP | 15. RING_FINGER_DIP |
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| 6. INDEX_FINGER_PIP | 17. PINKY_MCP |
| 7. INDEX_FINGER_DIP | 18. PINKY_PIP |
| 8. INDEX_FINGER_TIP | 19. PINKY_DIP |
| 9. MIDDLE_FINGER_MCP | 20. PINKY_TIP |
| 10. MIDDLE_FINGER_PIP | |

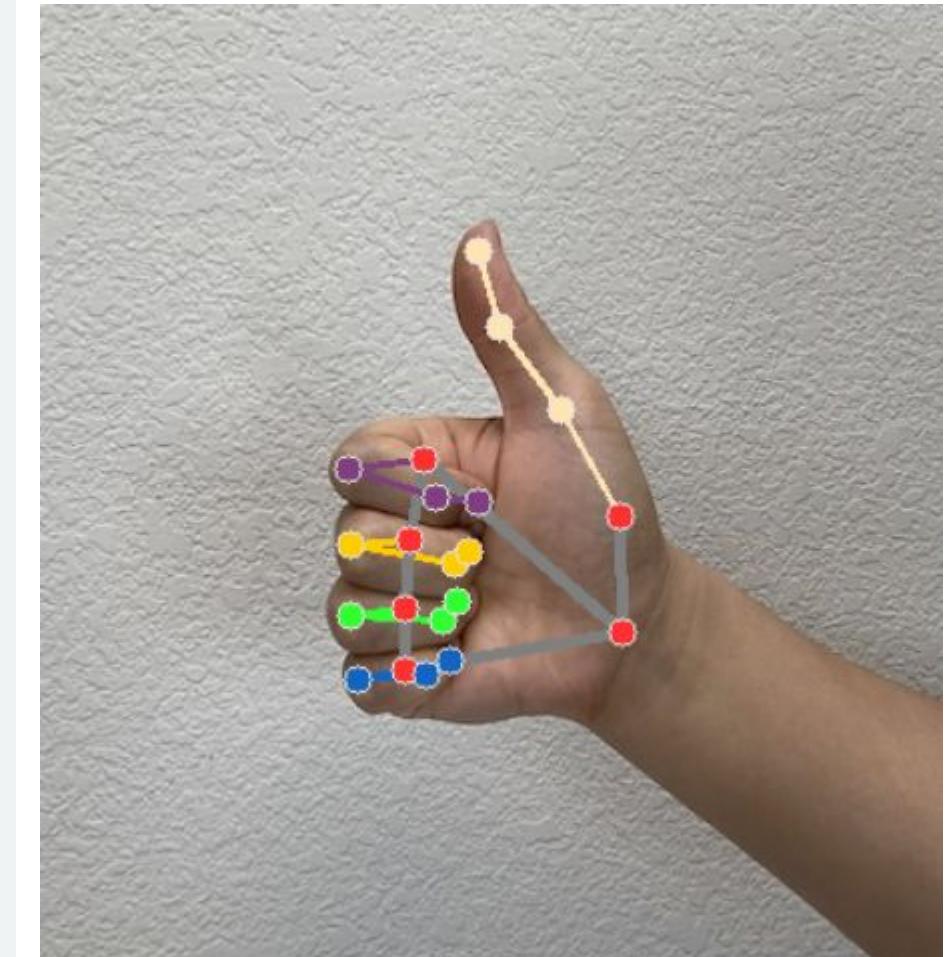


Hand Gestures Recognition

➤ What Is Hand Gestures Recognition?



```
GestureRecognizerResult:  
    Handedness:  
        Categories #0:  
            index      : 0  
            score      : 0.98396  
            categoryName : Left  
    Gestures:  
        Categories #0:  
            score      : 0.76893  
            categoryName : Thumb_Up  
    Landmarks:  
        Landmark #0:  
            x          : 0.638852  
            y          : 0.671197  
            z          : -3.41E-7  
        Landmark #1:  
            x          : 0.634599  
            y          : 0.536441  
            z          : -0.06984  
            ... (21 landmarks for a hand)  
    WorldLandmarks:  
        Landmark #0:  
            x          : 0.067485  
            y          : 0.031084  
            z          : 0.055223  
        Landmark #1:  
            x          : 0.063209  
            y          : -0.00382  
            z          : 0.020920  
            ... (21 world landmarks for a hand)
```

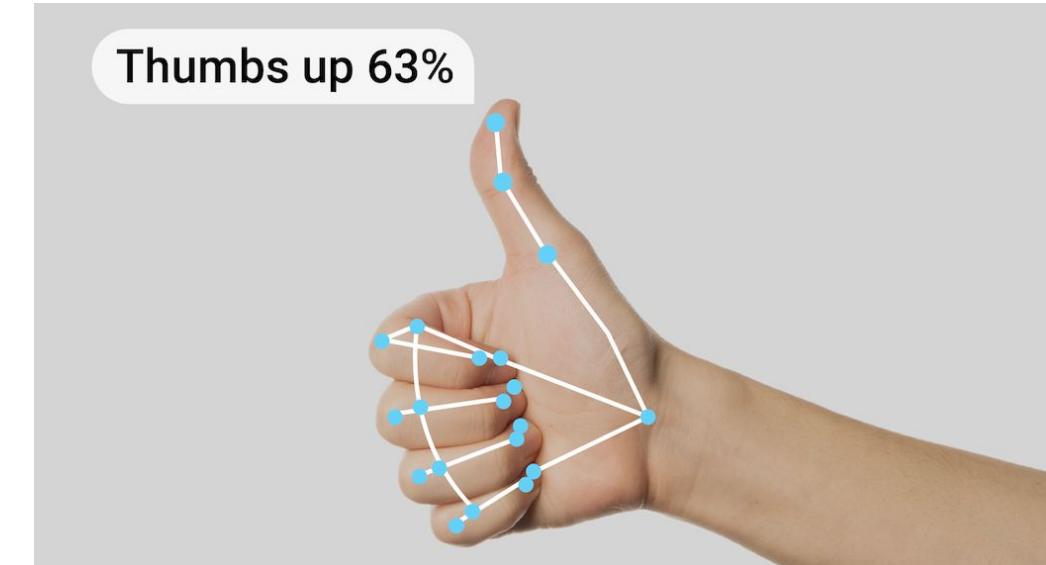


Hand Gestures Recognition

➤ What Is Hand Gestures Recognition?

Hand Gesture classification model: Hand gesture classification involves the recognition of specific hand movements, which can be used in various applications such as human-computer interaction, sign language interpretation, and virtual reality. Several models and techniques have been developed to enhance the accuracy and efficiency of gesture recognition.

- 0 - Unrecognized gesture, label: Unknown
- 1 - Closed fist, label: Closed_Fist
- 2 - Open palm, label: Open_Palm
- 3 - Pointing up, label: Pointing_Up
- 4 - Thumbs down, label: Thumb_Down
- 5 - Thumbs up, label: Thumb_Up
- 6 - Victory, label: Victory
- 7 - Love, label: ILoveYou

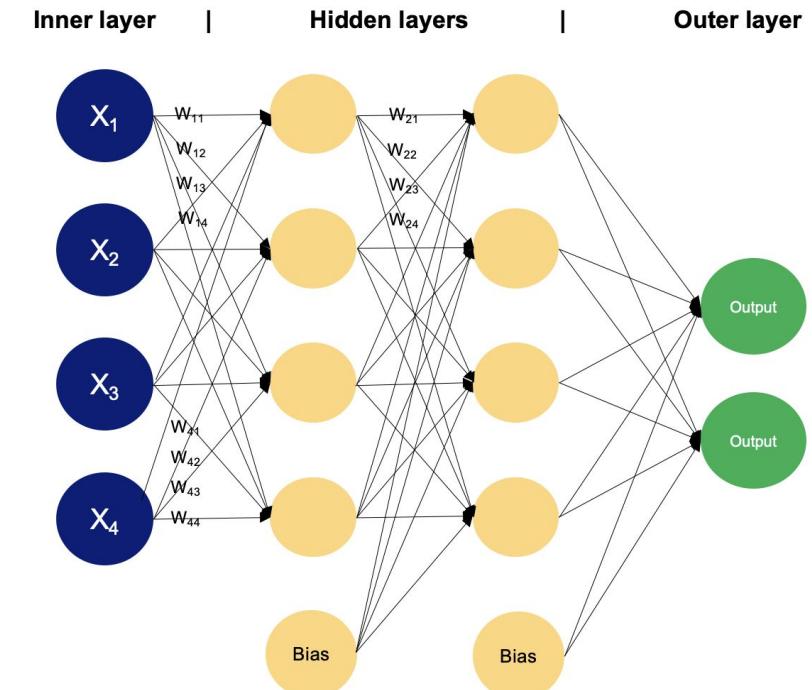
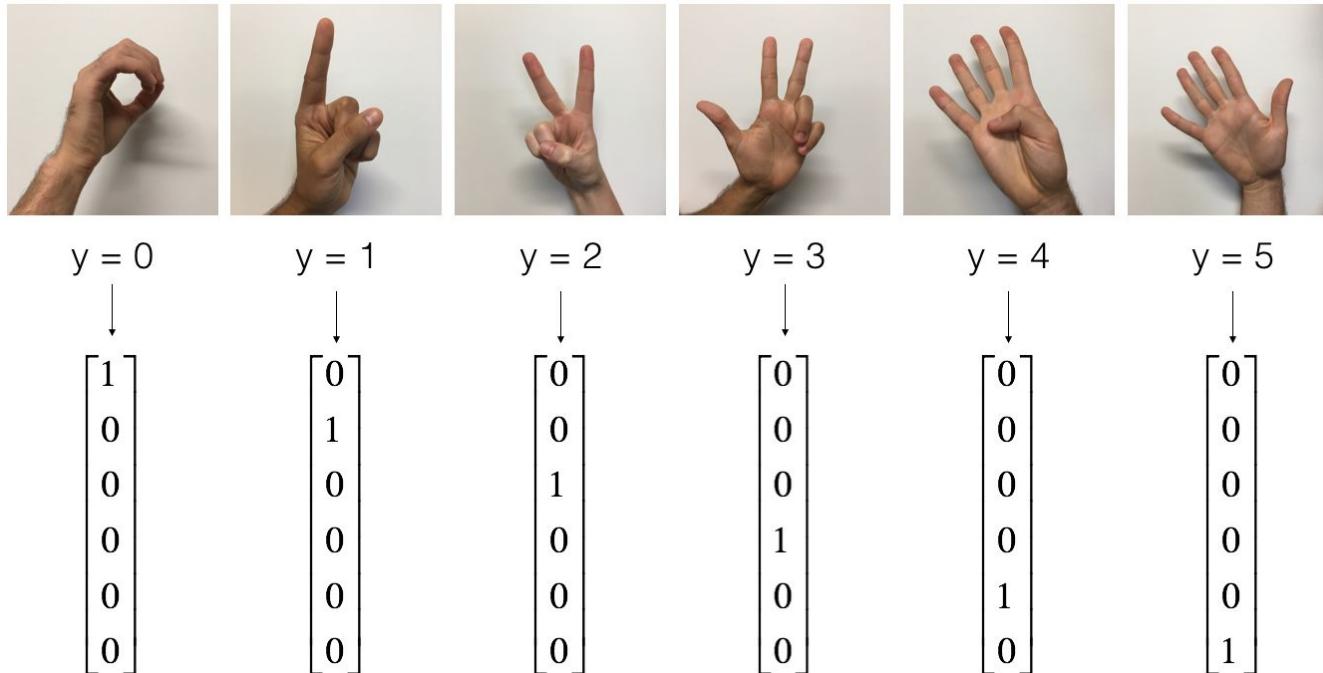


Hand Gestures Recognition

➤ What Is Hand Gestures Recognition?

MLP Model Structure

- Input Layer: Receives features extracted from gesture data
- Hidden Layers: One or more layers that process the inputs through weighted connections and nonlinear activation functions.
- Output Layer: Produces the final classification output, indicating the recognized gesture.

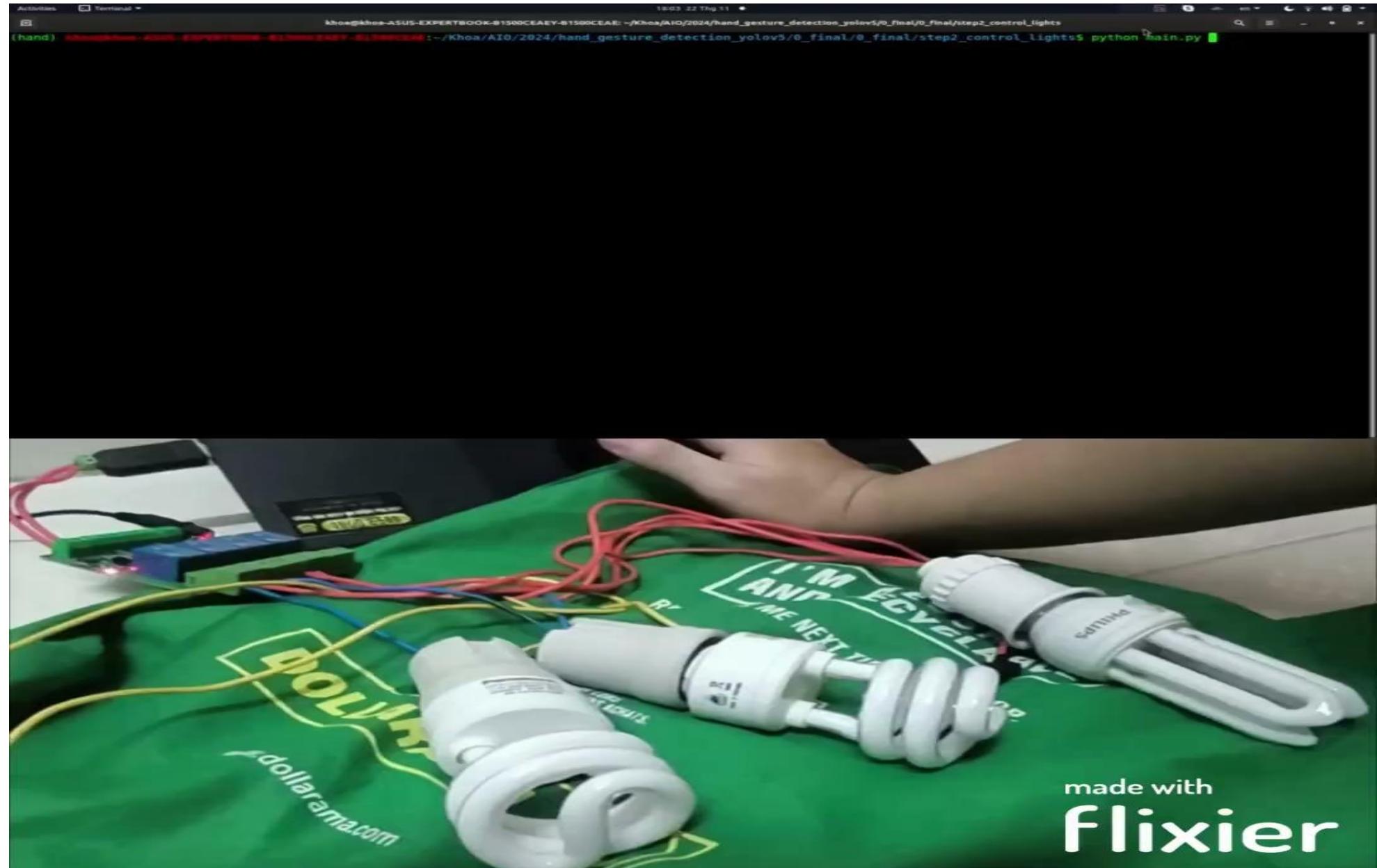


Project Overview

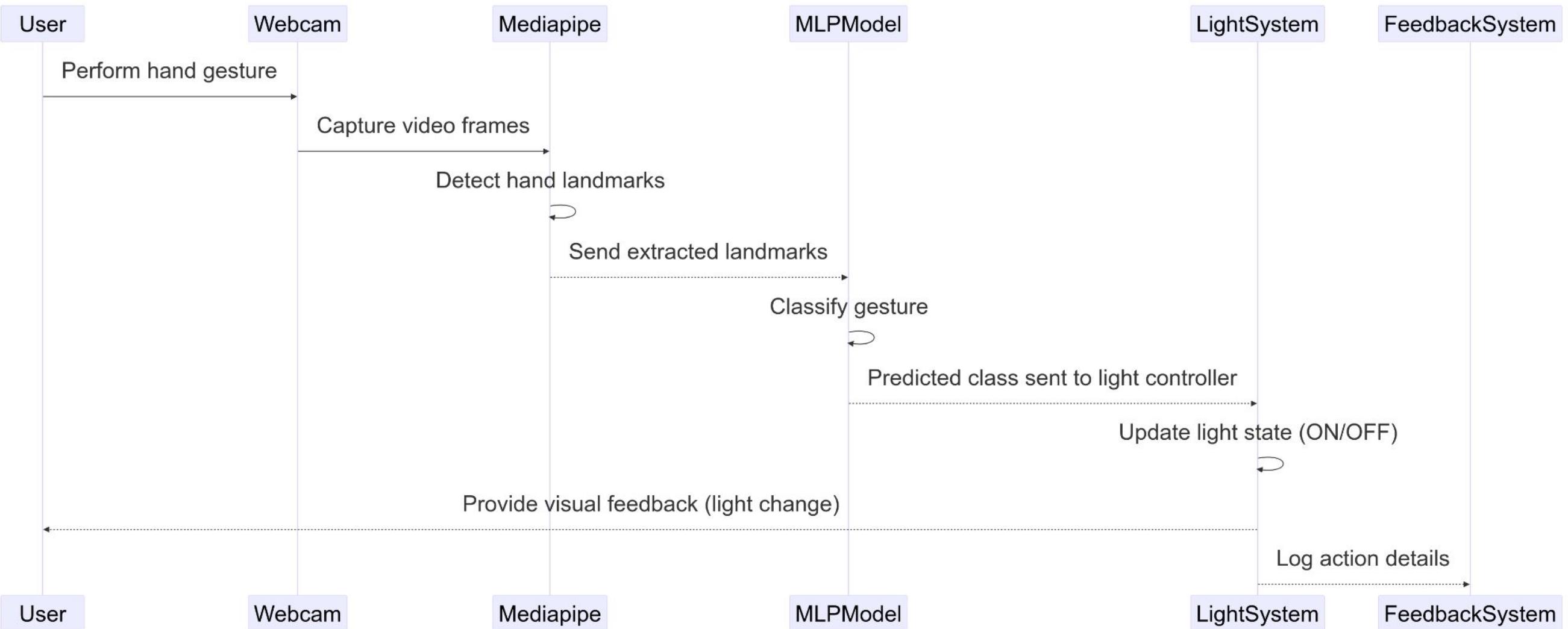
➤ Demo

```
step2_control_lights/  
└── controller.py  
└── detect_simulation.py  
└── hand_gesture.yaml
```

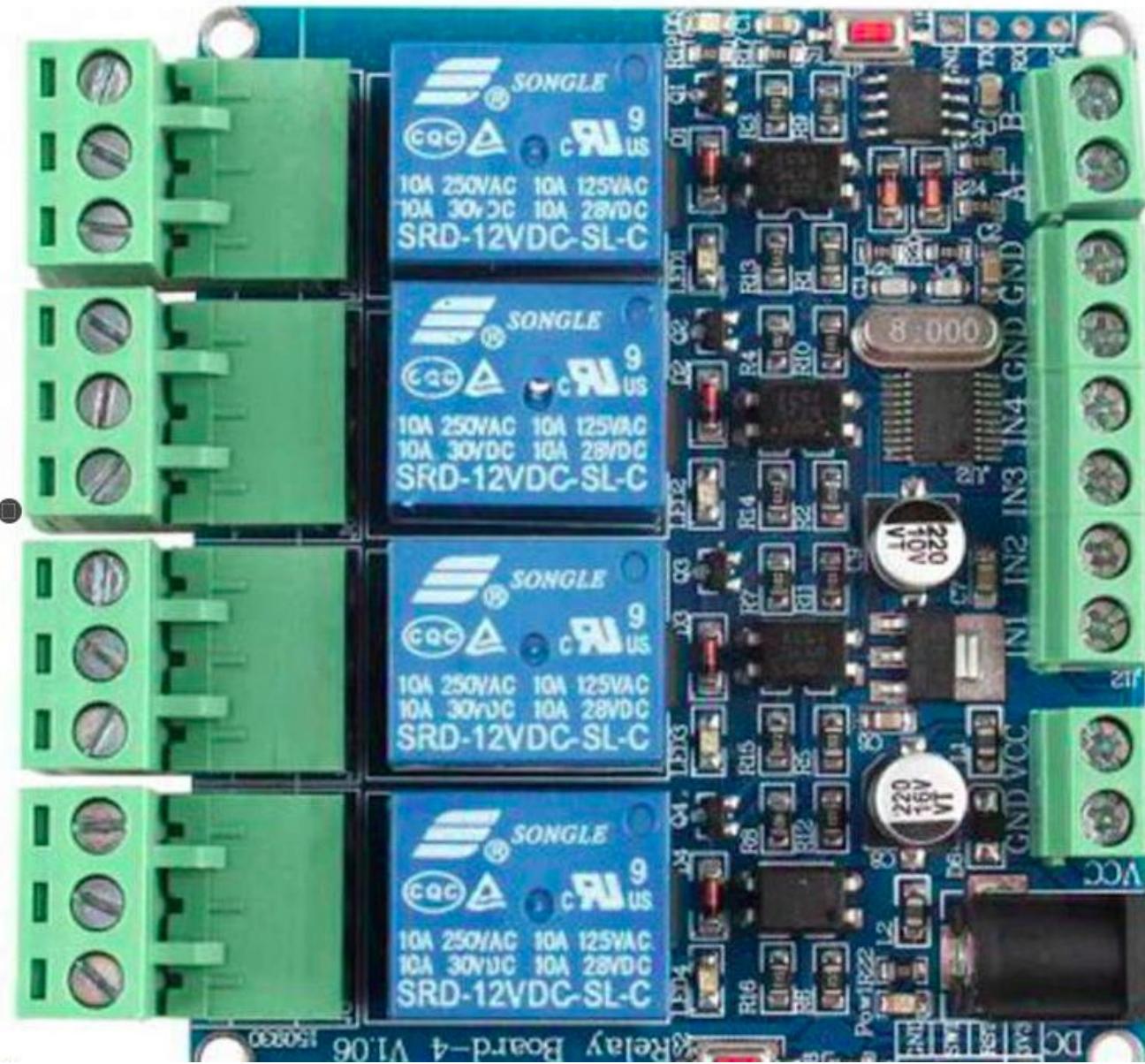
là file main.py
trong clip demo



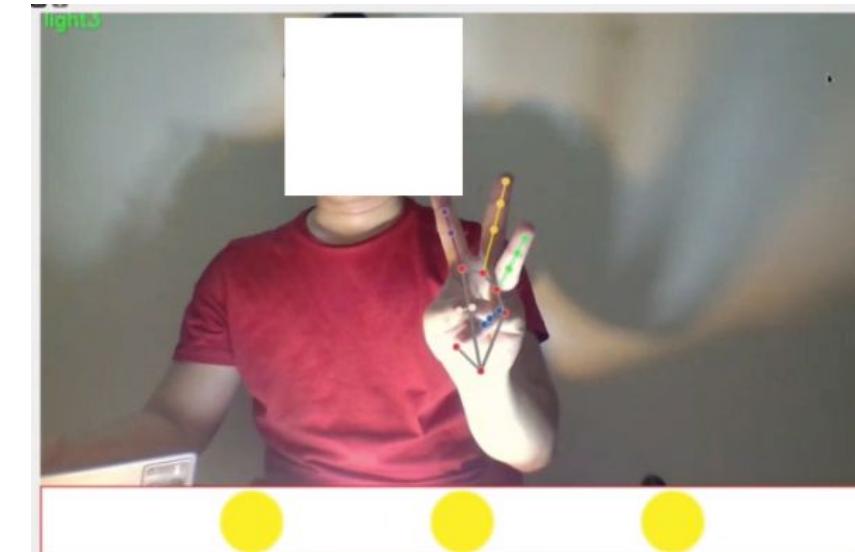
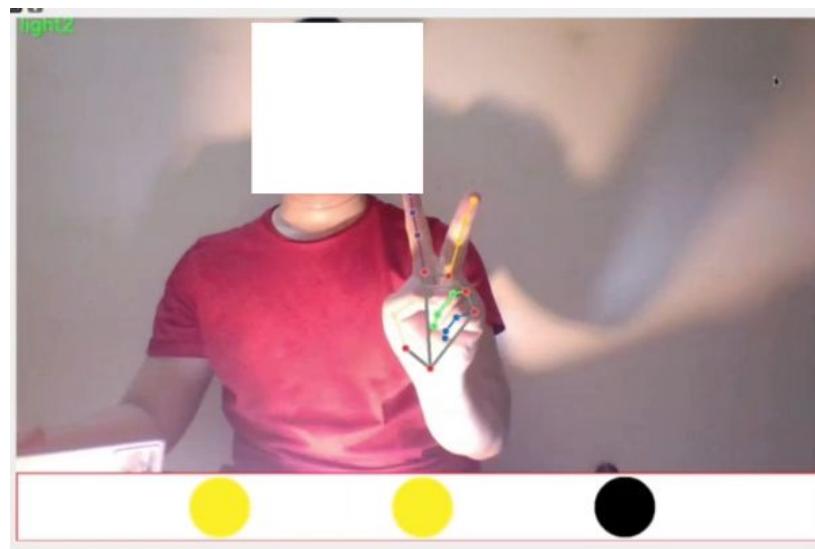
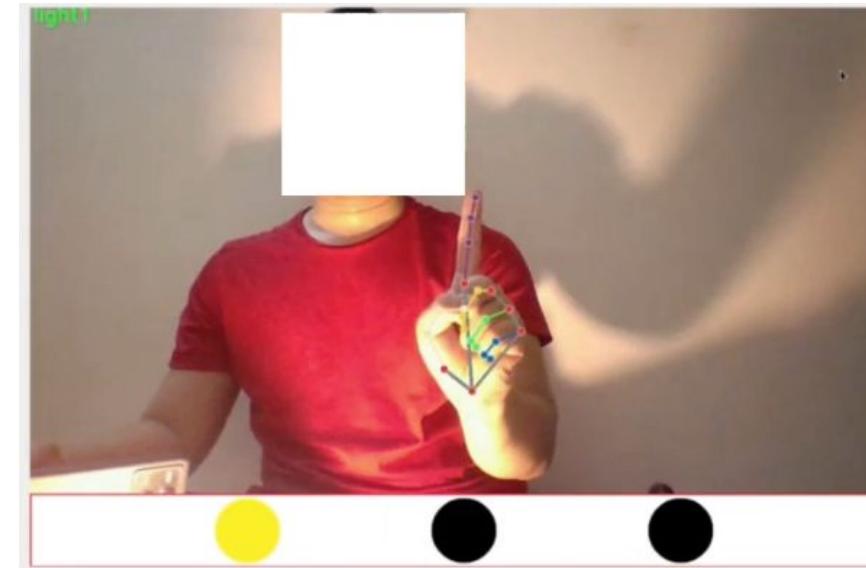
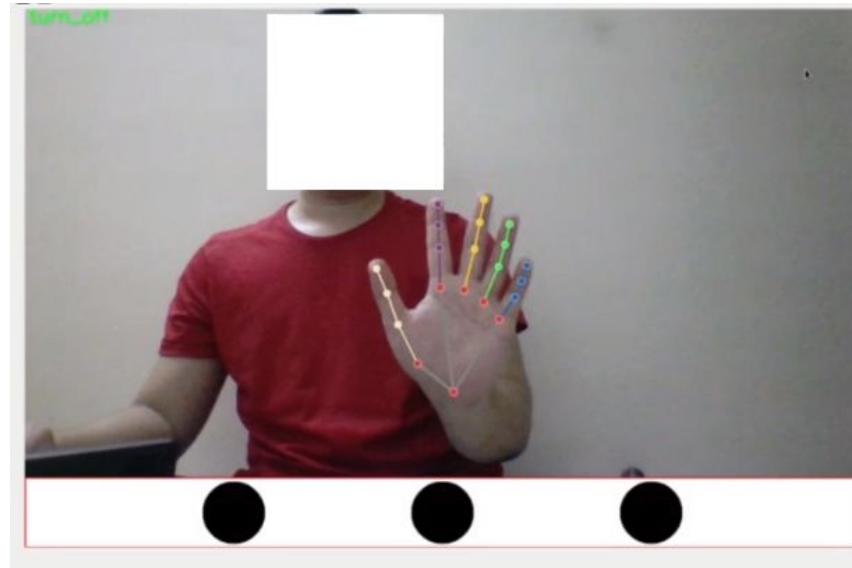
Project Overview



Project Overview



Project Overview



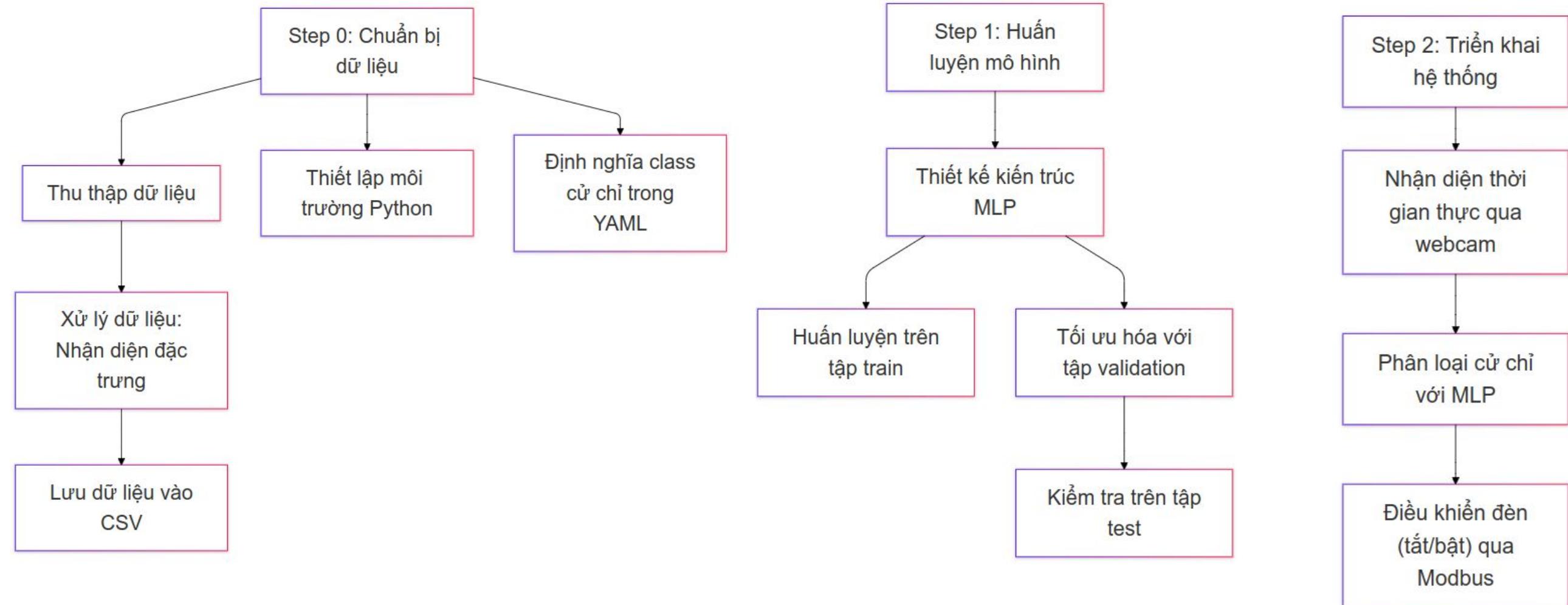
Quiz

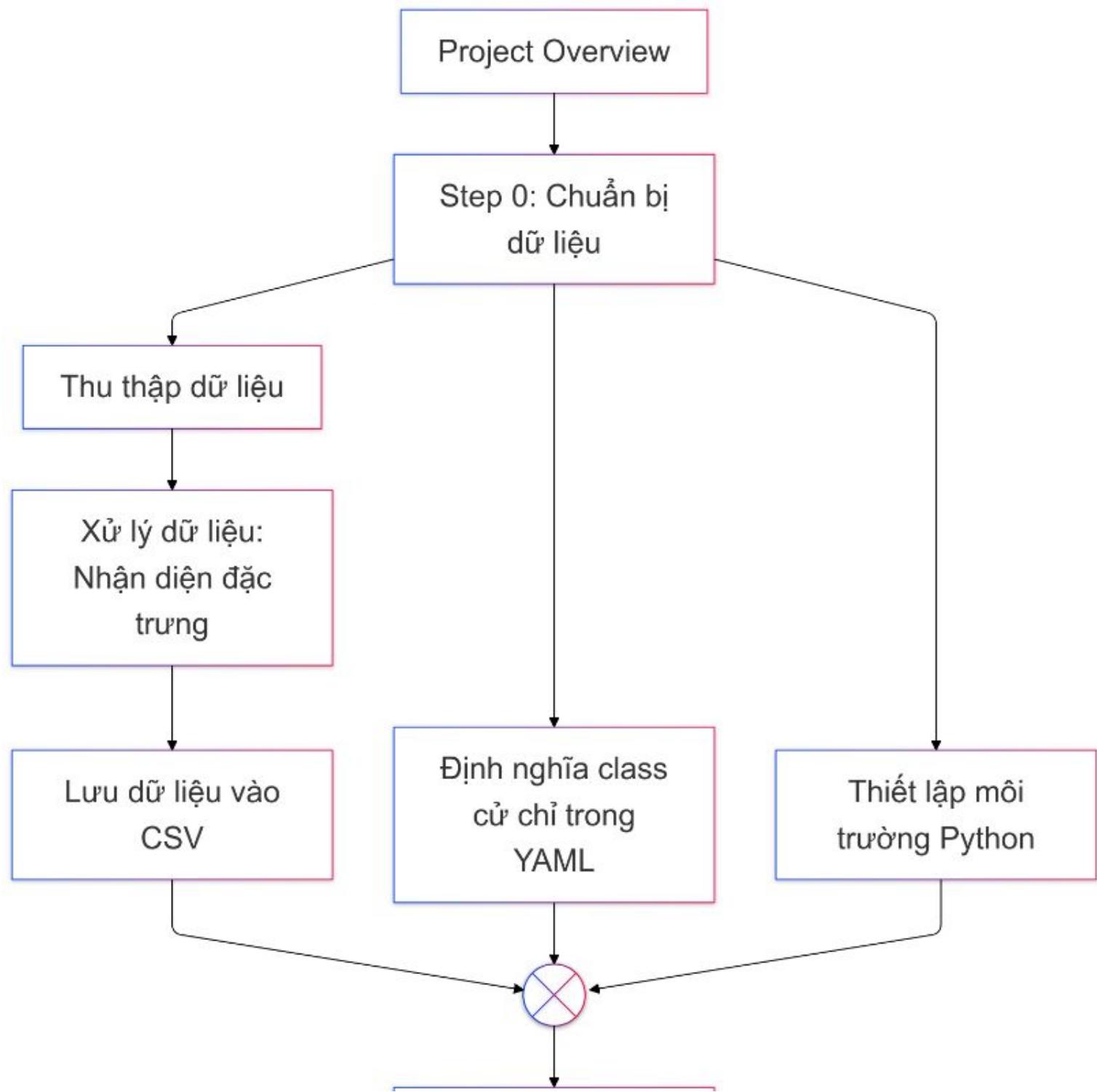
Thư viện nào được sử dụng để detect hand landmarks

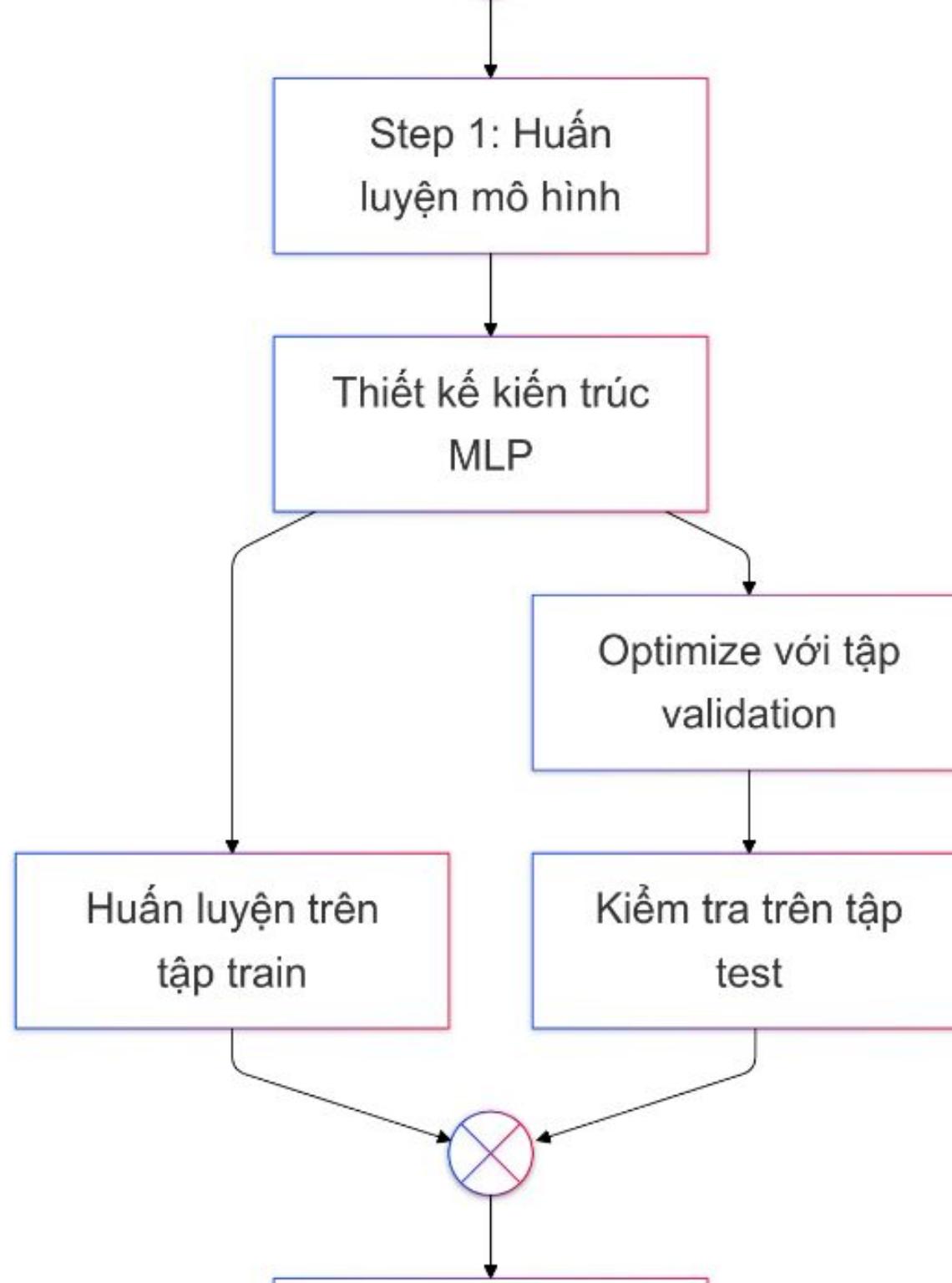
- (A) . Pytorch
- (B). Tensorflow
- (C). Scikit Learn
- (D). Mediapipe

Project Overview

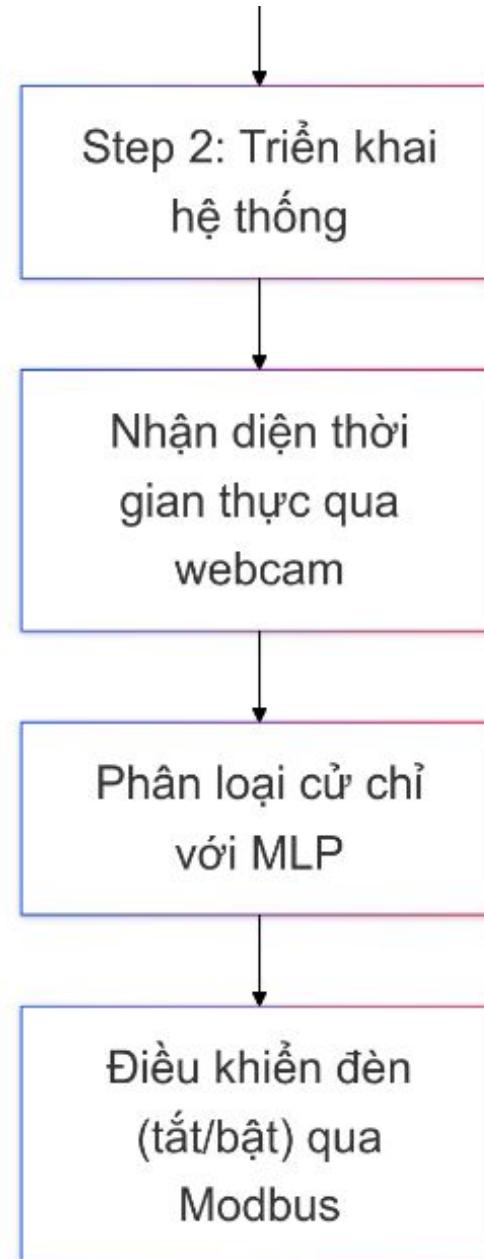
➤ Steps





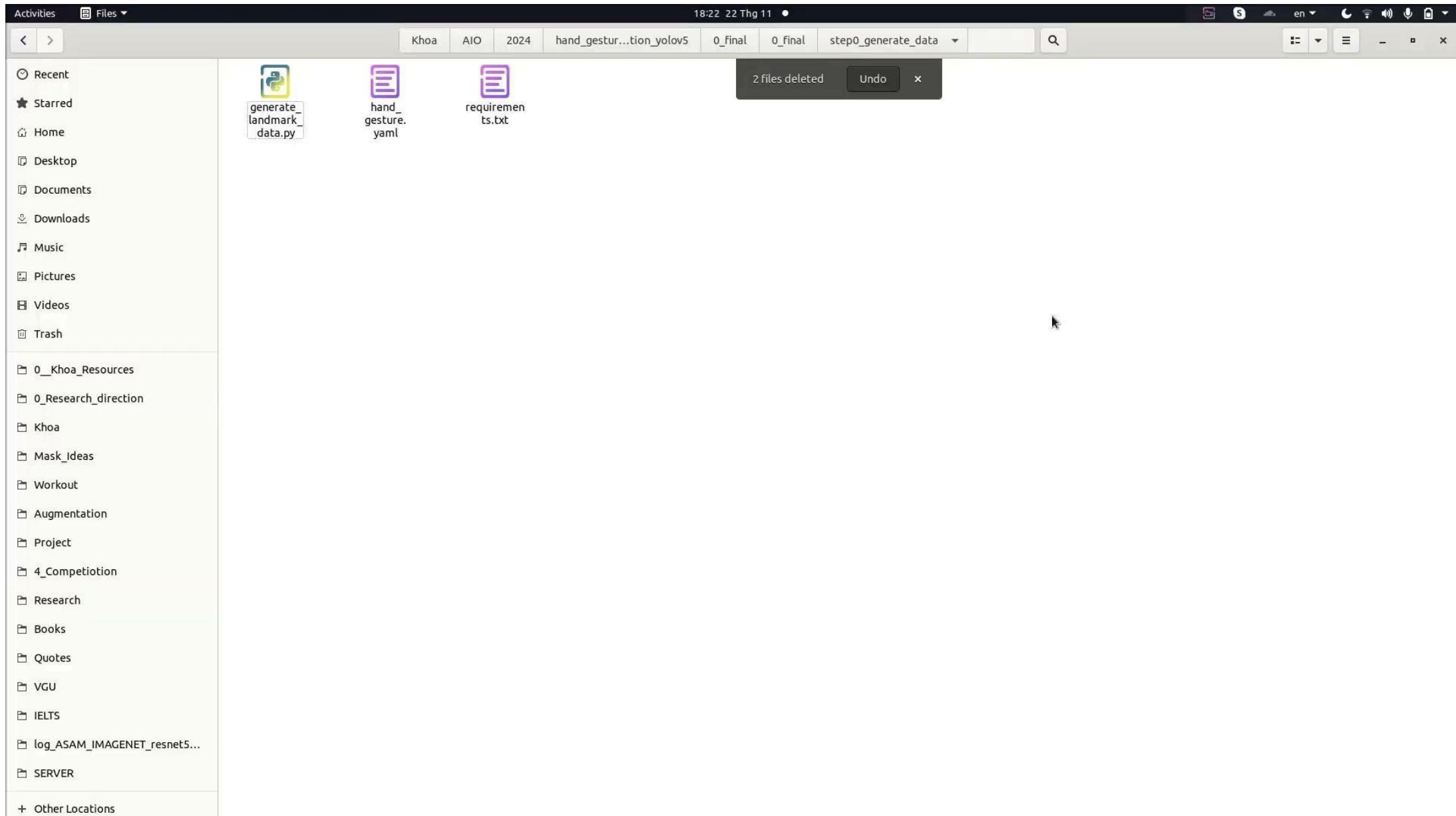


Project Overview



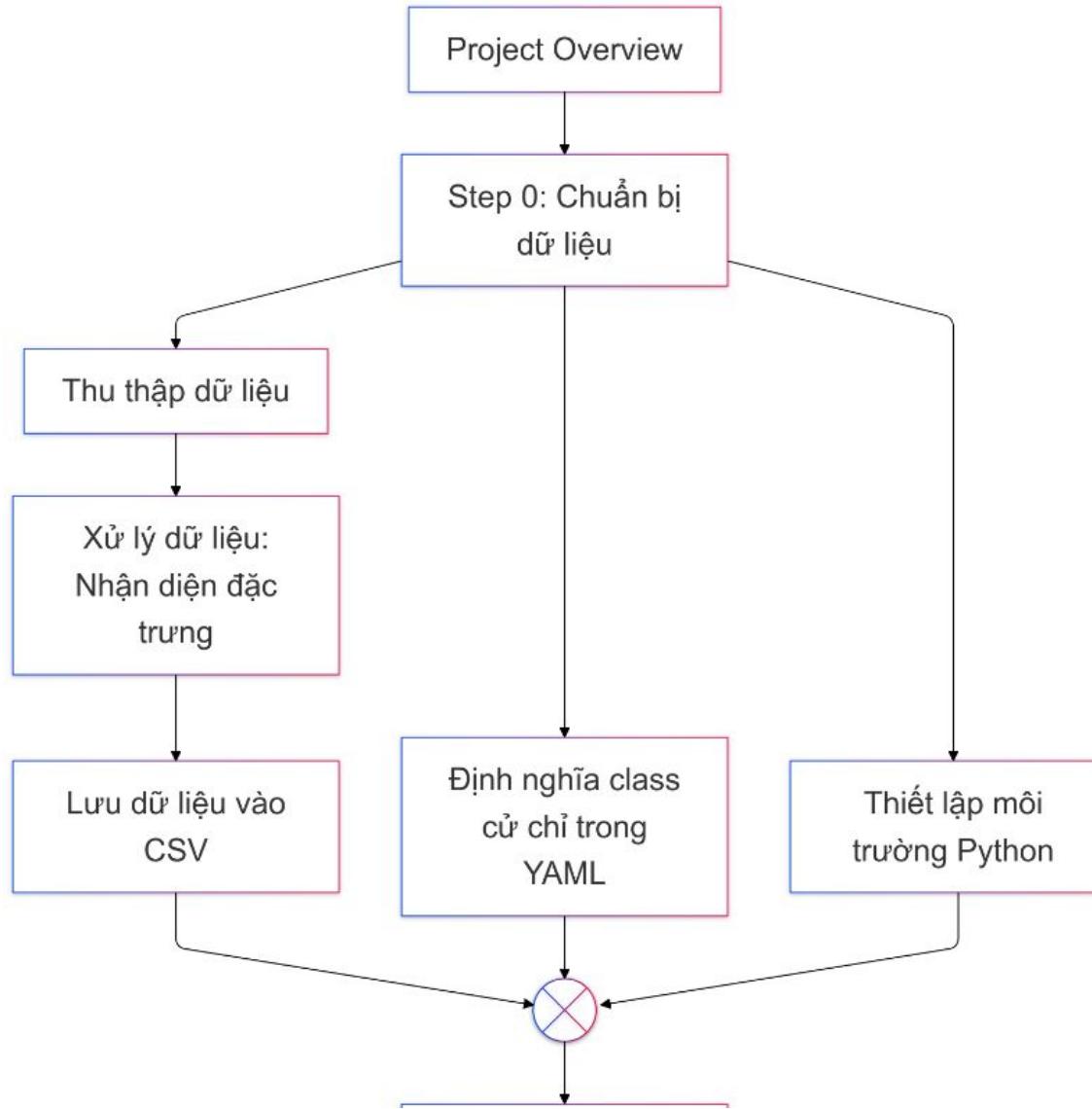
Step 0: Generate Data

➤ Generate Data



Step 0: Generate Data

➤ Generate Data

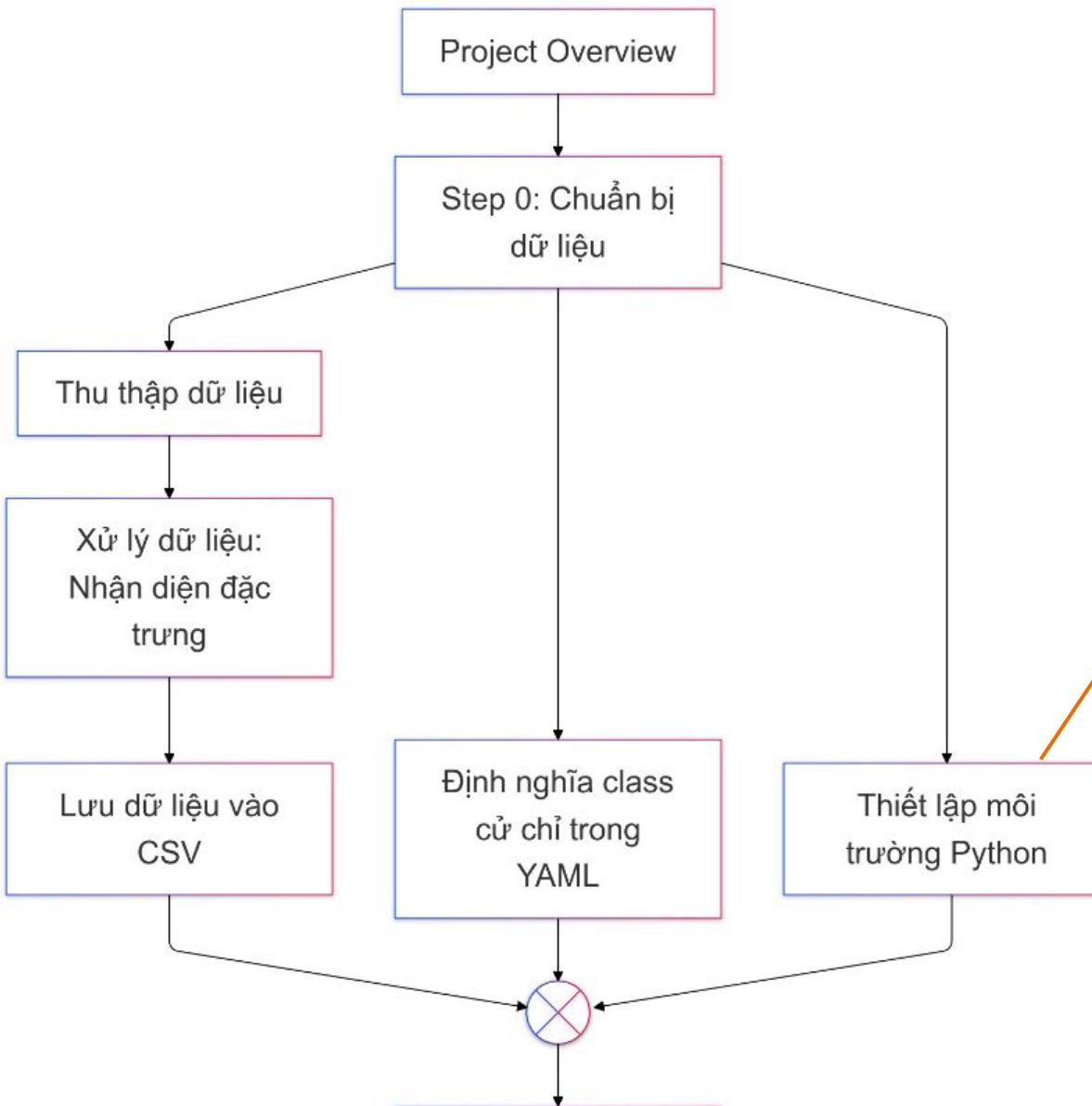


```
step0_generate_data
├── generate_landmark_data.py
└── hand_gesture.yaml
    └── requirements.txt
```

```
step0_generate_data
├── data
│   ├── landmark_test.csv
│   ├── landmark_train.csv
│   └── landmark_val.csv
│   └── generate_landmark_data.py
│   └── hand_gesture.yaml
│   └── requirements.txt
└── sign_imgs
    ├── light1.jpg
    ├── light2.jpg
    ├── light3.jpg
    └── turn_off.jpg
    └── turn_on.jpg
```

Step 0: Generate Data

➤ Generate Data



```
step0_generate_data
├── generate_landmark_data.py
└── hand_gesture.yaml
    └── requirements.txt
```

1. Thiết Lập Môi Trường Làm Việc

1.1. Sử Dụng Conda Với Python 3.10: Chúng ta sẽ sử dụng Conda để tạo môi trường ảo cho project với yêu cầu sử dụng Python phiên bản 3.10

- Cài đặt Anaconda hoặc Miniconda: Nếu bạn chưa có Conda trên máy tính, hãy tải và cài đặt từ trang chủ:

- Anaconda: <https://www.anaconda.com/download>
- Miniconda: <https://docs.anaconda.com/miniconda/>

- Tạo môi trường mới với Python 3.10:

```
1 conda create -n gesture_env python=3.10.0
```

- Kích hoạt môi trường:

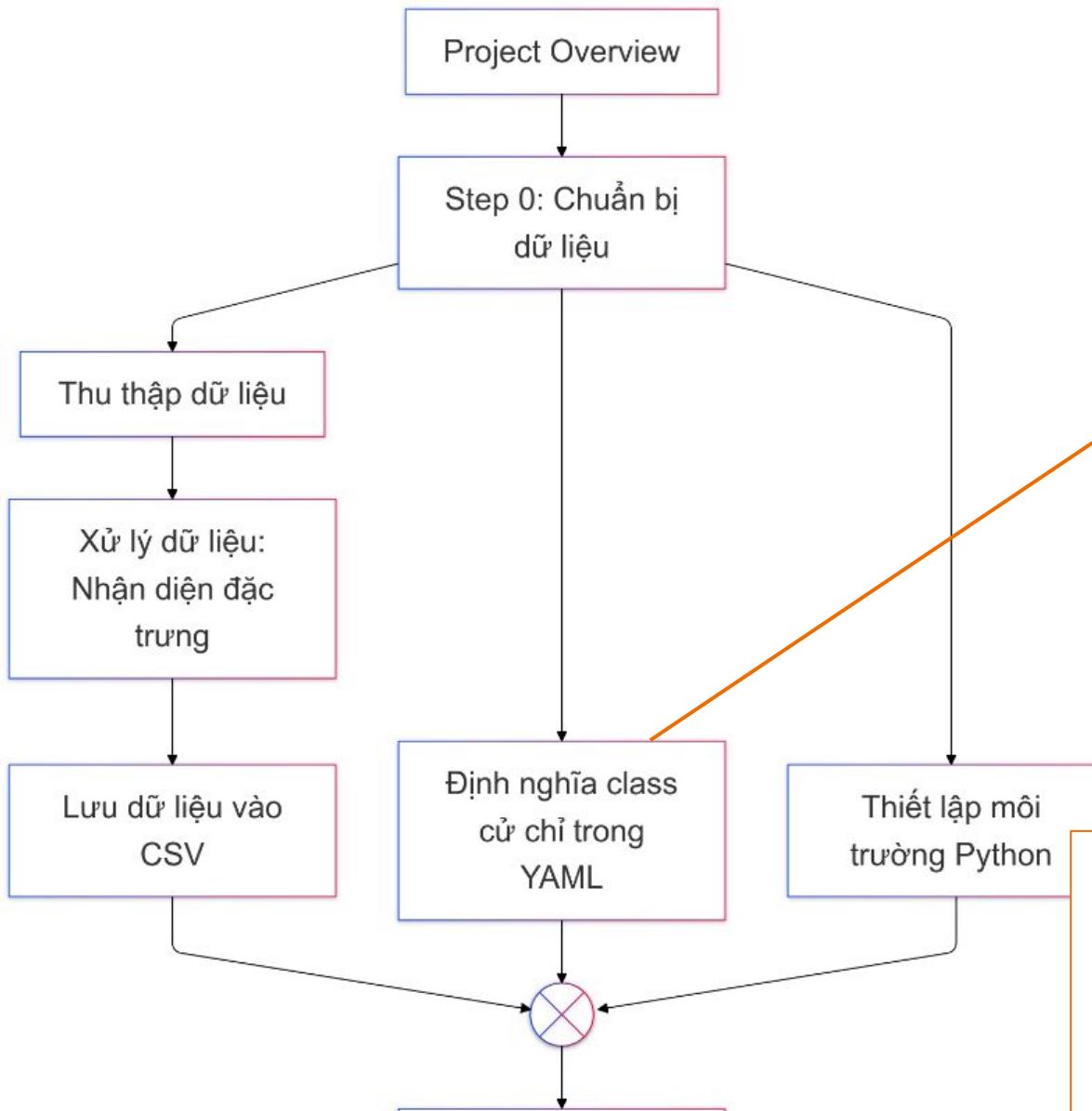
```
1 conda activate gesture_env
```

1.2. Cài Đặt Các Thư Viện Từ requirements.txt: Sau khi môi trường được activate, chúng ta tiến hành cài đặt các thư viện cần thiết được liệt kê trong file requirements.txt.

```
1 pip install -r requirements.txt
```

Step 0: Generate Data

➤ Generate Data



```
step0_generate_data
├── generate_landmark_data.py
└── hand_gesture.yaml
    └── requirements.txt
```

```
1 gestures:
2   0: "turn_off"
3   1: "light1" #
4   2: "light2" #
5   3: "light3" #
6   4: "turn_on"
```

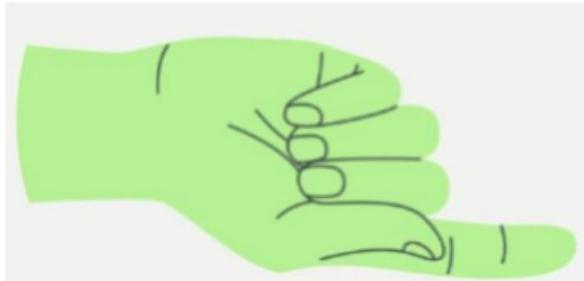
Giải thích:

- **gestures:** Đây là key, chứa danh sách các cù chỉ.
- 0, 1, 2, 3, 4: Các số đại diện cho class của từng loại cù chỉ.
- "turn_off", "light1", "light2", "light3", "turn_on": Tên của các cù chỉ, đồng thời cũng là hành động điều khiển đèn tương ứng.

Step 0: Generate Data

➤ Generate Data

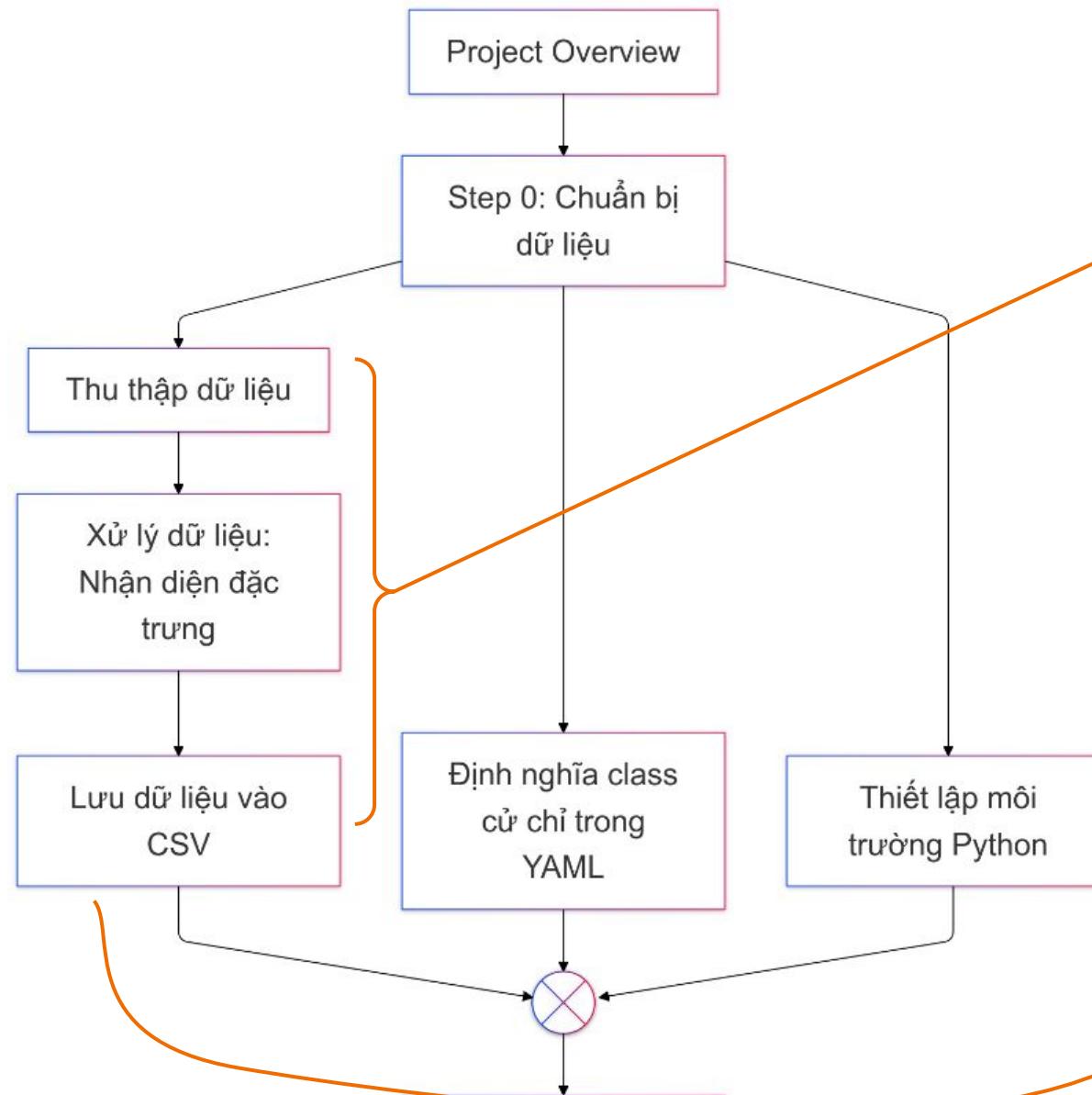
```
1 gestures:  
2   0: "turn_off"  
3   1: "light1" #  
4   2: "light2" #  
5   3: "light3" #  
6   4: "turn_on"
```

Class	Action	Hand Gesture
0	Turn All the Lights Off	
1	Turn the Lights 1 On	

2	Turn the Lights 2 On	
3	Turn the Lights 3 On	
4	Turn All the Lights On	

Step 0: Generate Data

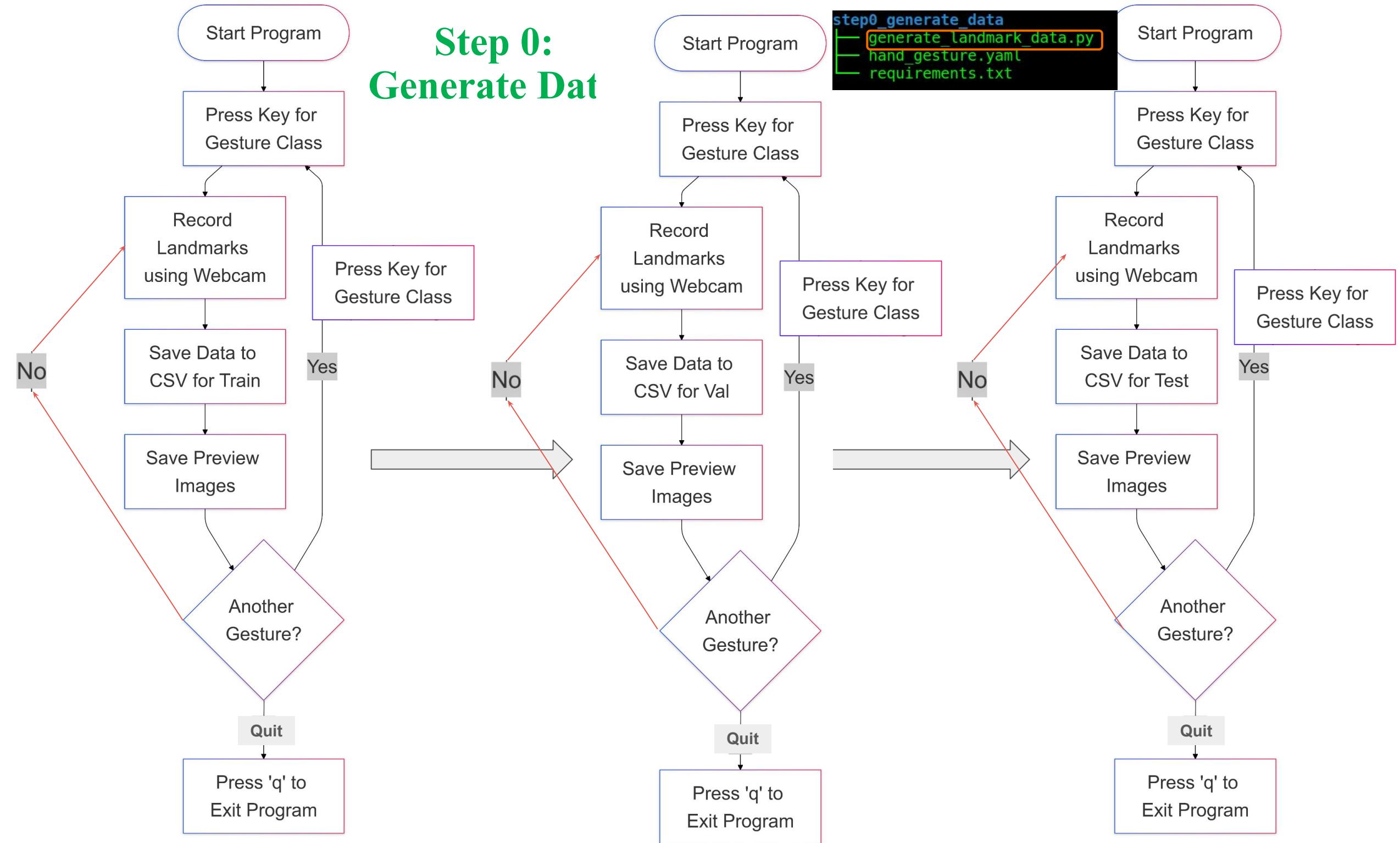
➤ Generate Data



The diagram shows two file structures:

- step0_generate_data**:
 - generate_landmark_data.py
 - hand_gesture.yaml
 - requirements.txt
- step0_generate_data**:
 - data**:
 - landmark_test.csv
 - landmark_train.csv
 - landmark_val.csv
 - generate_landmark_data.py
 - hand_gesture.yaml
 - requirements.txt
 - sign_imgs**:
 - light1.jpg
 - light2.jpg
 - light3.jpg
 - turn_off.jpg
 - turn_on.jpg

Step 0: Generate Dat



Step 0: Generate Data

step0_generate_data
└── generate_landmark_data.py
└── hand_gesture.yaml
└── requirements.txt

➤ Generate Data

```
step0_generate_data
├── data
│   ├── landmark_test.csv
│   ├── landmark_train.csv
│   └── landmark_val.csv
├── generate_landmark_data.py
├── hand_gesture.yaml
└── requirements.txt
└── sign_imgs
    ├── light1.jpg
    ├── light2.jpg
    ├── light3.jpg
    ├── turn_off.jpg
    └── turn_on.jpg
```

Sử dụng cho step 1

Kiểm tra các cử chỉ có chính xác

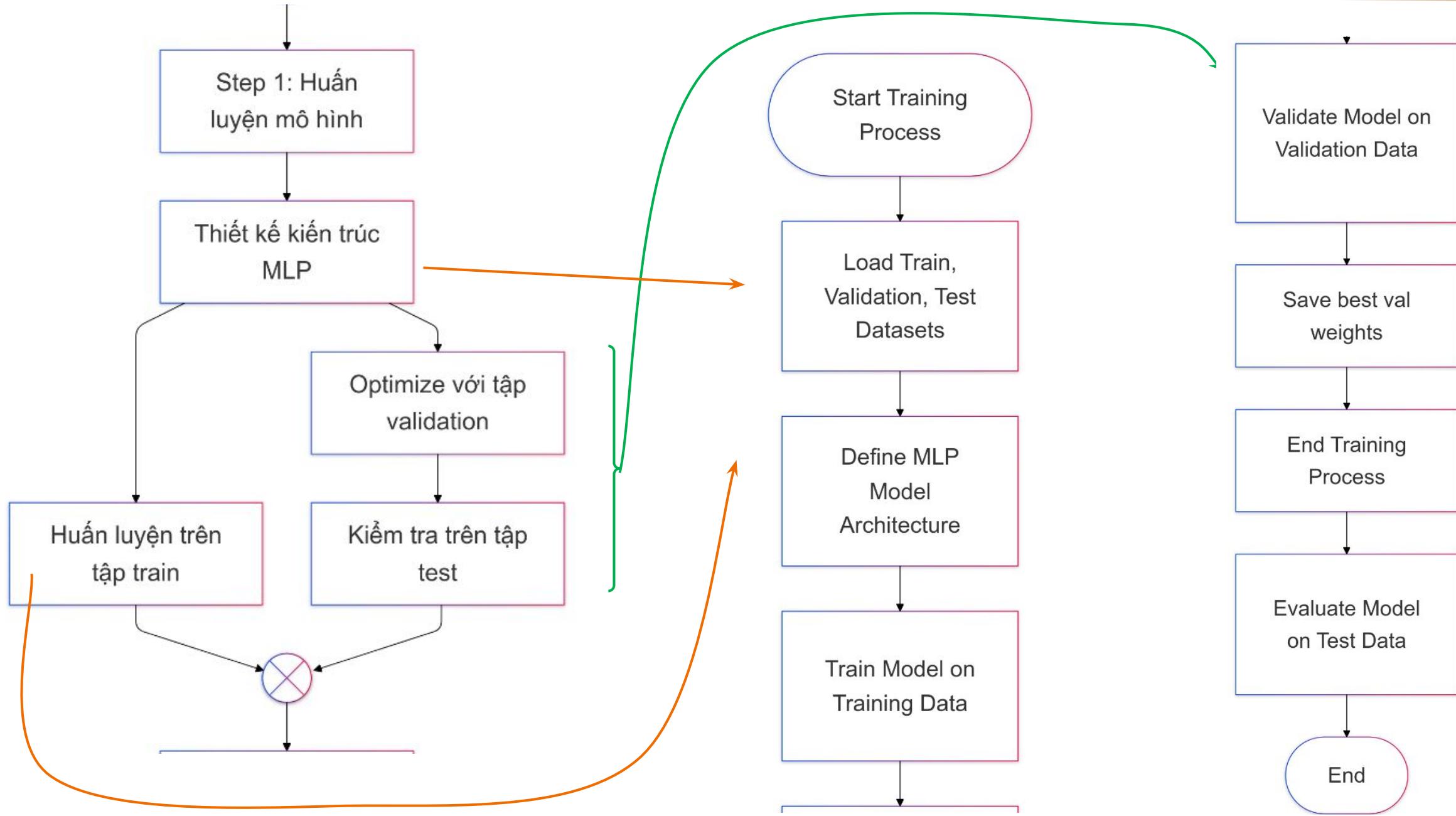


Quiz

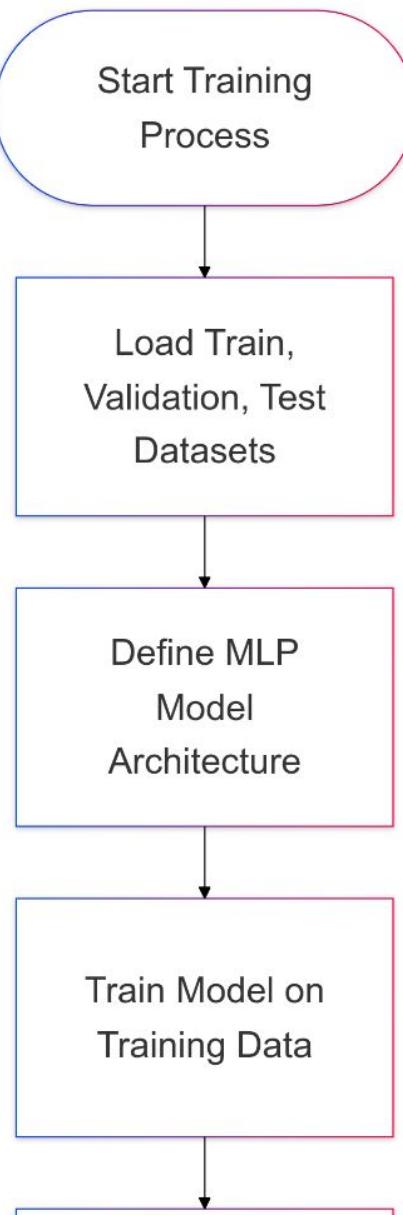
Tại Step 0 file nào định nghĩa các class cù chỉ để điều khiển đèn?

- (A) . requirements.txt
- (B). generate_landmark_data.py
- (C). hand_gesture.yaml
- (D). gesture_classes.csv

Step 1: Hand Gesture



Step 1: Hand Gesture



```
1 DATA_FOLDER_PATH="../data/"
2 LIST_LABEL = label_dict_from_config_file("hand_gesture.yaml")
3 train_path = os.path.join(DATA_FOLDER_PATH,"landmark_train.csv")
4 val_path = os.path.join(DATA_FOLDER_PATH,"landmark_val.csv")
5 save_path = './models'
6 os.makedirs(save_path,exist_ok=True)
7
8 trainset = CustomImageDataset(train_path)
9 ##### Your Code Here #####
10 '''Hoàn thành code để thực hiện khởi tạo DataLoader cho trainset với
11 batch_size 40 và cho phép xáo trộn
12 '''
13 trainloader =
14 #####
15
16 valset = CustomImageDataset(os.path.join(val_path))
17 val_loader = torch.utils.data.DataLoader(valset,batch_size=50, shuffle=False)
```

Step 1: Hand Gesture

Start Training
Process

Load Train,
Validation, Test
Datasets

Define MLP
Model
Architecture

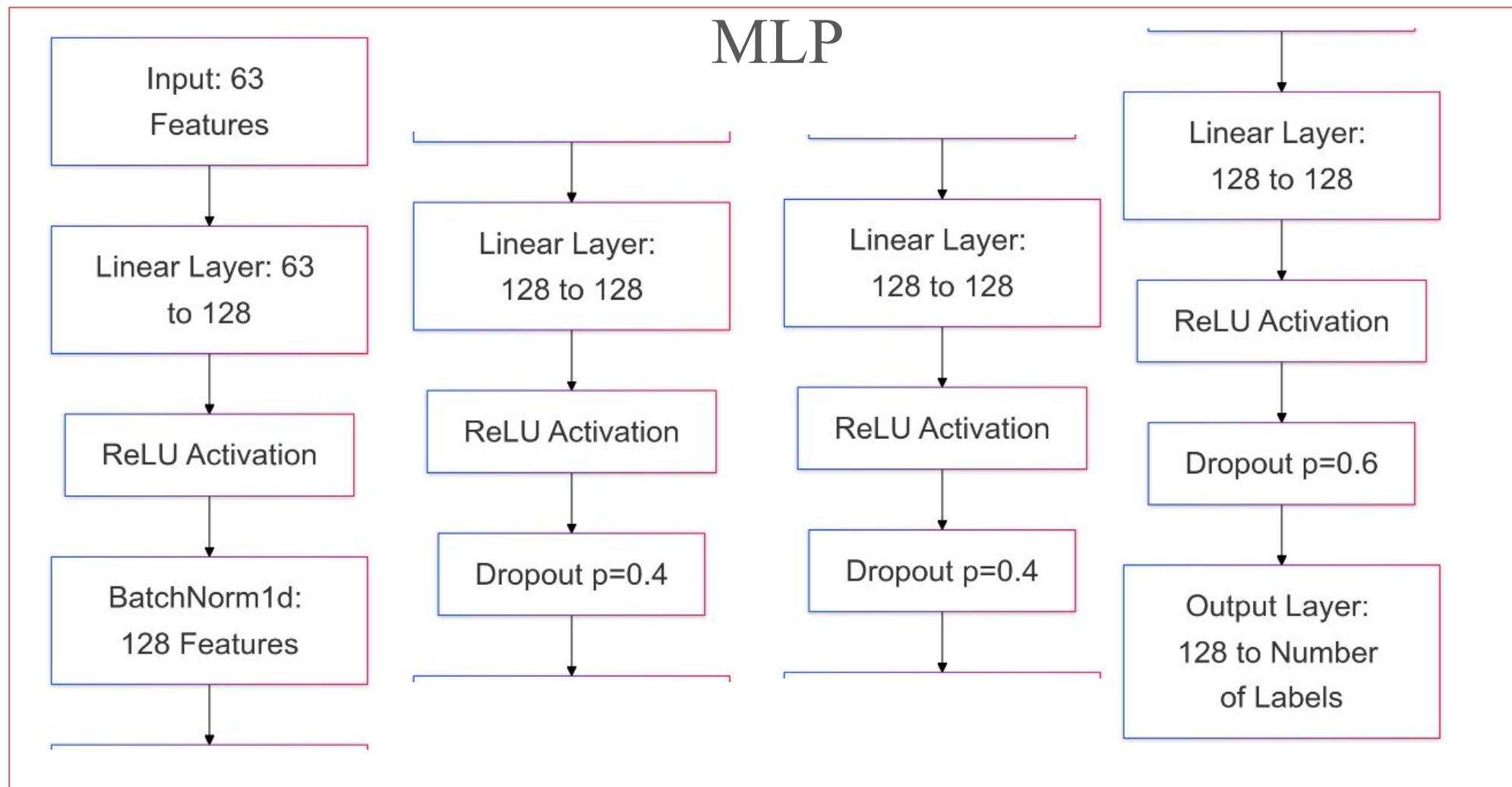
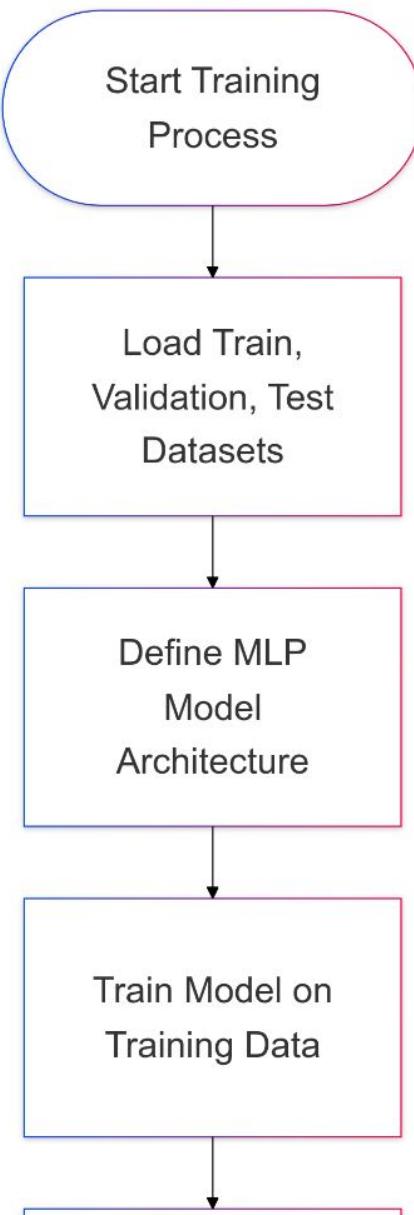
Train Model on
Training Data

```
1 # Tạo folder data để các bạn upload file 3 file data được tạo ra ở step 0 vào folder data
2 # Các bạn upload file hand_gesture.yaml ngoài folder data
3 !mkdir data/
4
5
6
7
8
9
10
11
12
13
14
15
16
```

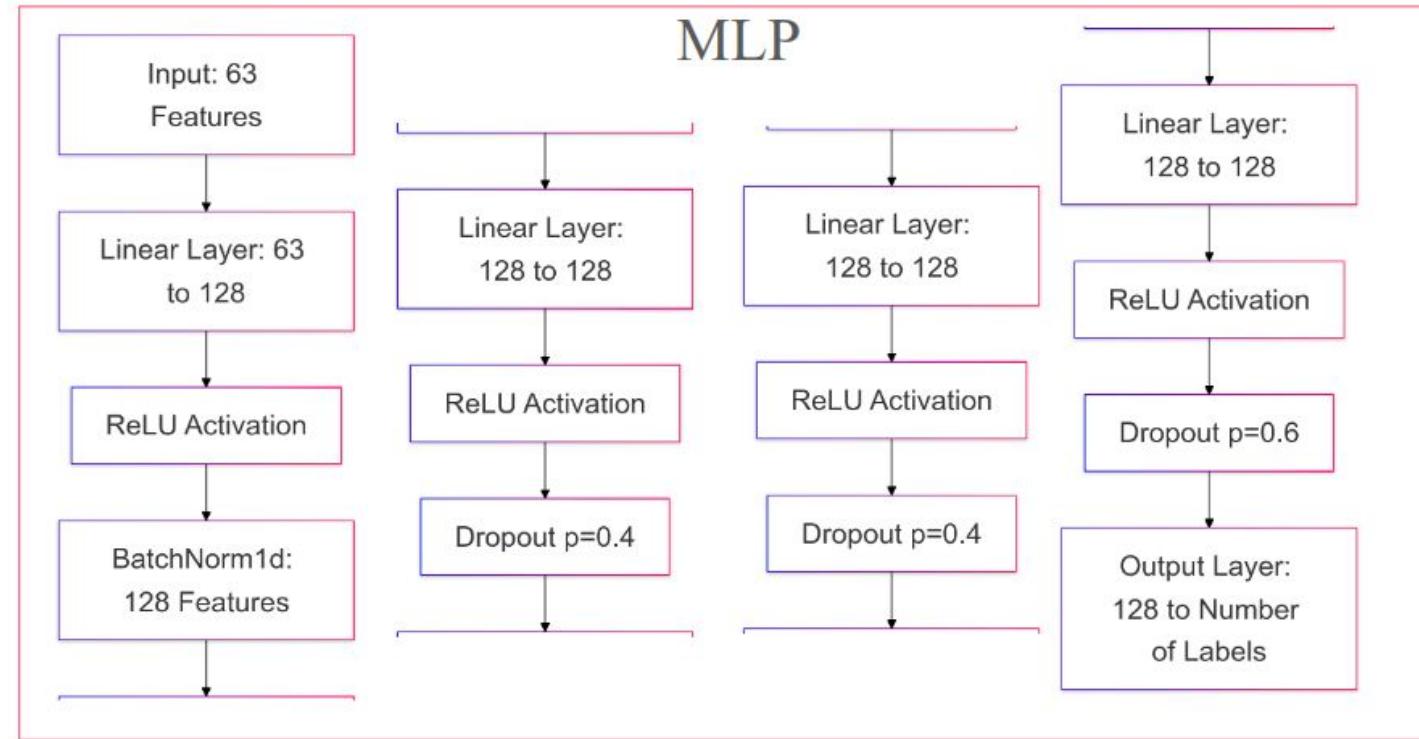
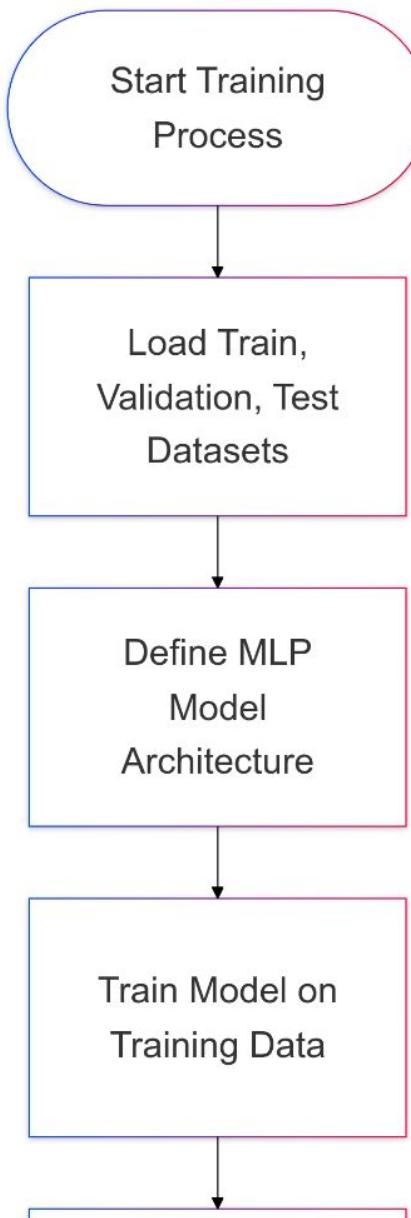
Nếu đã chạy thành công step 0 thì các bạn không chạy cell này và chỉ upload 3 file data của các bạn vào thư mục data
Và upload file hand_gesture.yaml bên ngoài thư mục data
Nếu không chạy được ở step 0 thì các bạn run cell này để download data
!gdown 1gzW0tABiVmJ38usCSDe5F9gR2tECt3zu -O data/
!gdown 15lwipssmC_K82ukRfb0uVCiDH1TZ3QCf -O data/
!gdown 1nIol_wBmkovz-u_BCsV5c1Kbz6ZqoKwq -O data/
Download file hand_gesture.yaml
!gdown 1ZteHYSgbuZu_GcUJHW8ZzoZv1DE8-oLw

```
1 class NeuralNetwork(nn.Module):
2     def __init__(self):
3         super(NeuralNetwork, self).__init__()
4         self.flatten = nn.Flatten()
5         list_label = label_dict_from_config_file("hand_gesture.yaml")
6
7         self.linear_relu_stack = nn.Sequential(
8             ##### Your Code Here #####
9             '''Hoàn thành đoạn code để xây dựng một model gồm có 4 hidden layer,
10            lân lượng input và output là (63, 128), (128, 128), (128, 128), (128, 128).
11            Layer đầu tiên được theo sau bởi một Relu và Batchnorm1d.
12            Layer thứ 2, 3, và 4 được theo sau bởi Relu và Dropout với rate lân lượt là 0.4, 0.4, 0.6.
13            Output layer có nhiệm vụ phân loại với input là 128 và output là số lượng class cù'chỉ
14            '''
15
16         )
```

Step 1: Hand Gesture

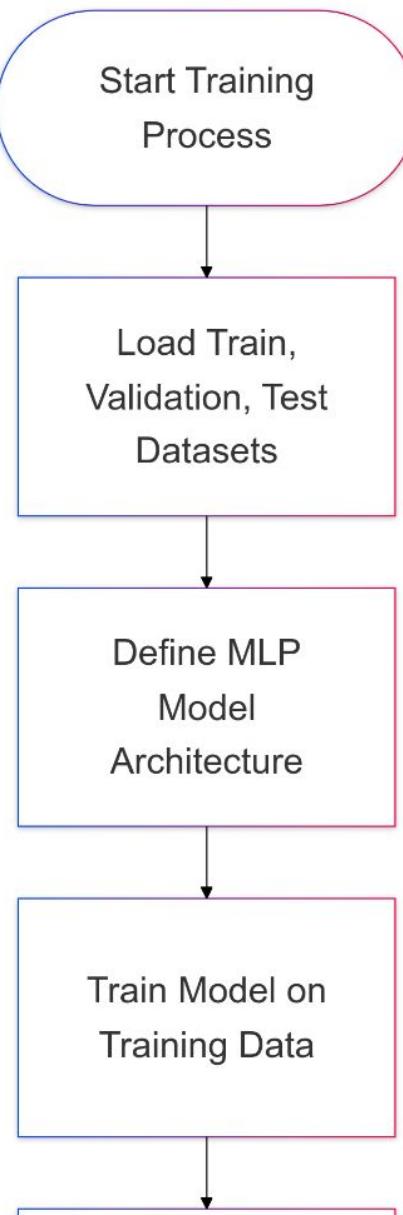


Step 1: Hand Gesture



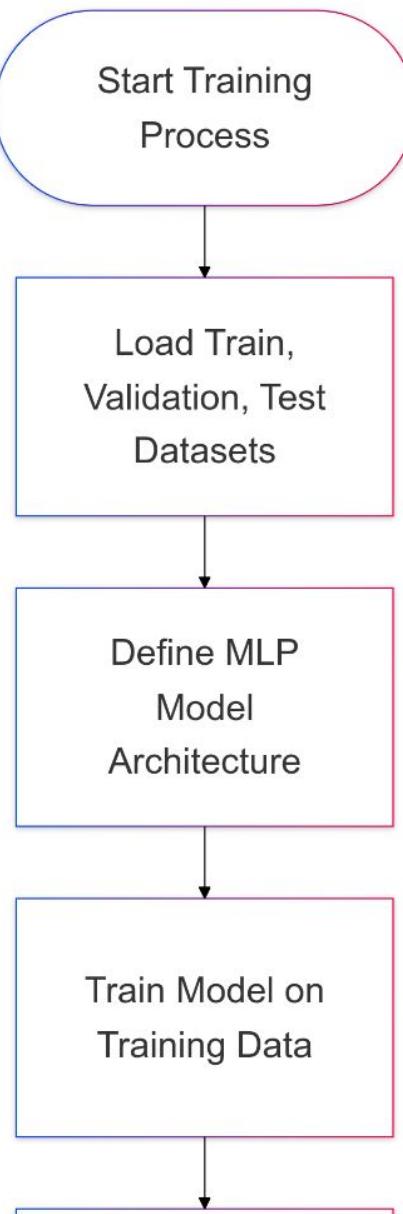
```
def forward(self, x):
    ##### Your Code Here #####
    ''' Hoàn thành code để thực hiện forward dự đoán cù'chỉ với input x.
    Thực hi'ent flatten x
    Pass x vừa flatten vào linear_relu_stack
    Return logits (outputs từ layer cuô' cùng)
    ...
    #####'''
```

Step 1: Hand Gesture



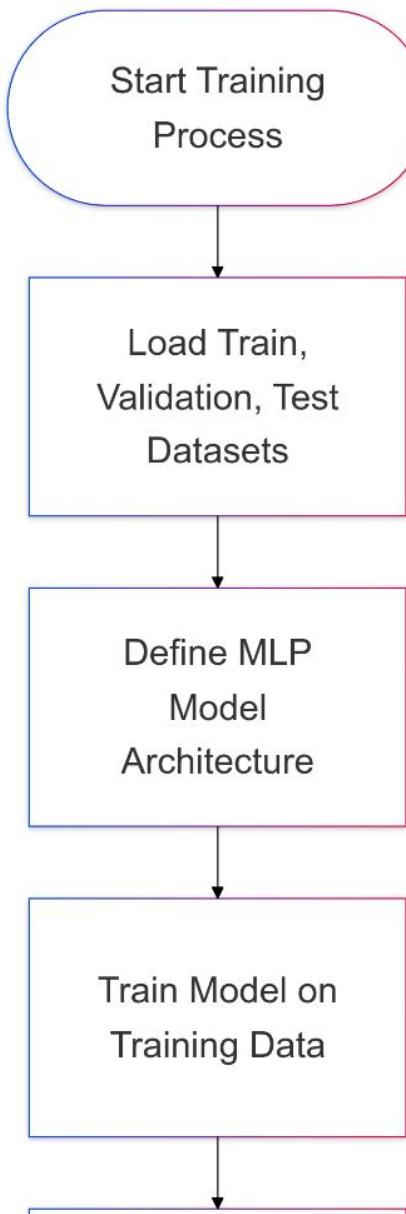
```
19 ##### Your Code Here #####
20 '''Hoàn thành code để thực hiện khởi tạo NeuralNetwork model đã xây dựng ở trên,
21 khởi tạo hàm loss sử dụng CrossEntropyLoss và khởi tạo early stopper với patience
22 là 30 và min_delta là 0.01
23 ...
24 model =
25 loss_function =
26 early_stopper =
27 #####
28
29 ##### Your Code Here #####
30 '''Hoàn thành code để thực hiện cấu hình Adam optimizer cho các tham số của
31 model với tốc độ học là 0.0001
32 ...
33 optimizer =
34 #####
35
36
37 model, best_model_path = train(trainloader, val_loader, model, loss_function, early_stopper, optimizer)
```

Step 1: Hand Gesture



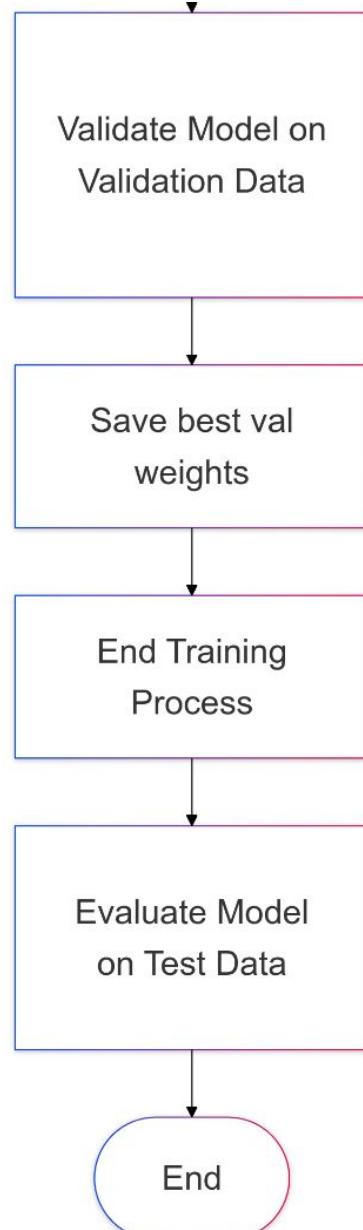
```
1 def train(trainloader, val_loader, model, loss_function, early_stopper, optimizer):
2     # add auroc score
3     best_vloss = 1_000_000
4     timestamp = datetime.now().strftime('%d-%m %H:%M')
5     for epoch in range(300):
6         #training step
7         model.train(True)
8         running_loss = 0.0
9         acc_train = Accuracy(num_classes=len(LIST_LABEL), task='MULTICLASS')
10        for batch_number,data in enumerate(trainloader):
11            inputs,labels = data
12
13            ##### Your Code Here #####
14            ''' Hoàn thành code để thực hiện reset gradients và dự đoán class của
15            chỉ của inputs
16            '''
17
18            preds =
19            #####
```

Step 1: Hand Gesture



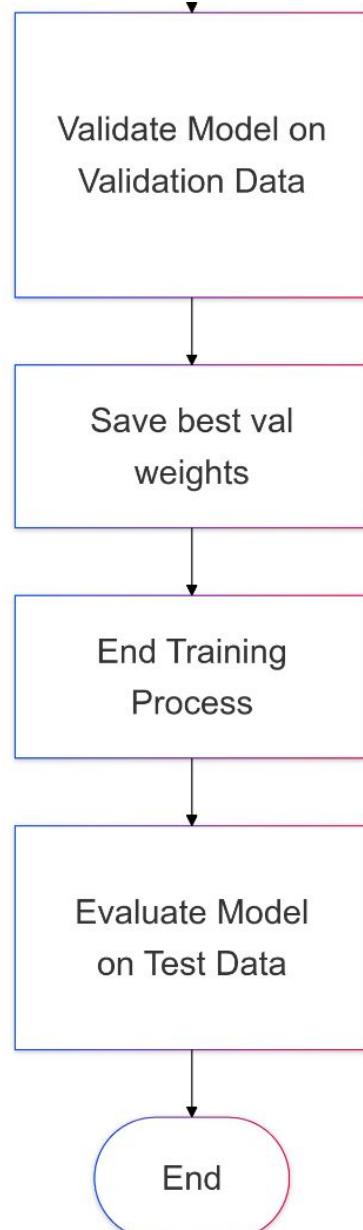
```
21 ##### Your Code Here ###### Q10
22
23     ''' Hoàn thành code để thực hiện tính loss dựa vào kết quả dự đoán
24     và labels, sau đó thực hiện backward và update parameters thông qua
25     optimizer
26     '''
27
28     loss =
29
30
31     acc_train.update(model.predict_with_known_class(inputs), labels)
32     running_loss += loss.item()
33     avg_loss = running_loss / len(trainloader)
```

Step 1: Hand Gesture



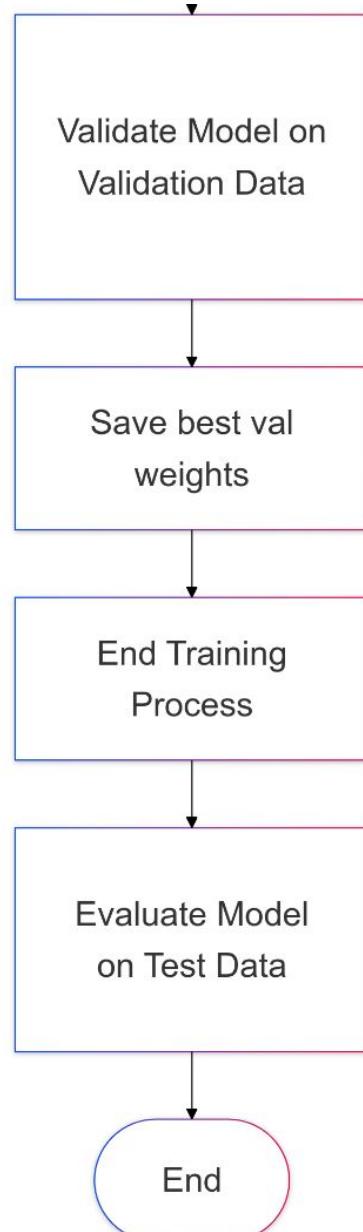
```
46     # Log the running loss averaged per batch
47     # for both training and validation
48     print(f"Epoch {epoch}: ")
49     print(f"Accuracy train:{acc_train.compute().item()}, val:{acc_val.compute().item()}")
50     avg_vloss = running_vloss / len(val_loader)
51     print('LOSS train {} valid {}'.format(avg_loss, avg_vloss))
52     print('Training vs. Validation Loss',
53           { 'Training' : avg_loss, 'Validation' : avg_vloss },
54           epoch + 1)
55     print('Training vs. Validation accuracy',
56           { 'Training' : acc_train.compute().item(),
57             'Validation' : acc_val.compute().item() },
58           epoch + 1)
59
60     # Track best performance, and save the model's state
61     if avg_vloss < best_vloss:
62         best_vloss = avg_vloss
63         best_model_path = f'./{save_path}/{model}_{timestamp}_{model.__class__.__name__}_best'
64         torch.save(model.state_dict(), best_model_path)
```

Step 1: Hand Gesture



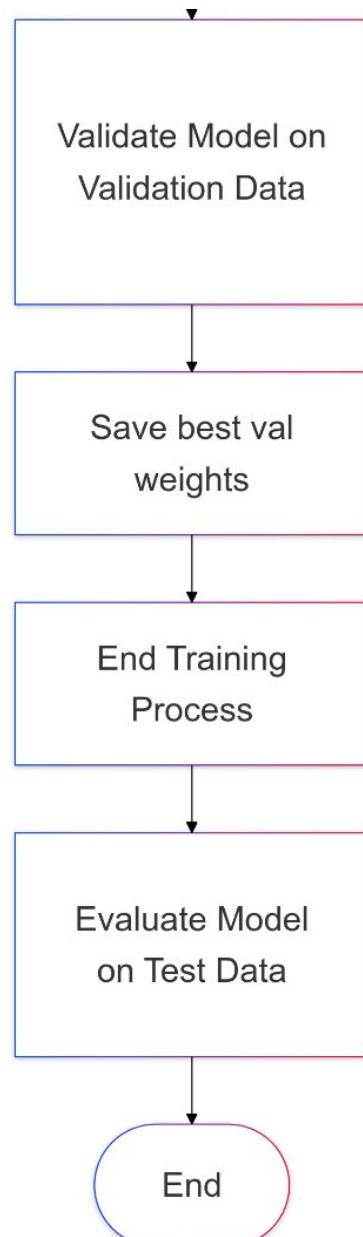
```
60 # Track best performance, and save the model's state
61 if avg_vloss < best_vloss:
62     best_vloss = avg_vloss
63     best_model_path = f'./{save_path}/model_{timestamp}_{model.__class__.__name__}_best'
64     torch.save(model.state_dict(), best_model_path)
65
66 if early_stopper.early_stop(avg_vloss):
67     ##### Your Code Here #####
68     ''' Hoàn thành đoạn code bên dưới để print ra epoch hiện tại và
69     minimum watched metric và thoát loop
70     '''
71
72 #####
73
74
75
76 model_path = f'./{save_path}/model_{timestamp}_{model.__class__.__name__}_last'
77 torch.save(model.state_dict(), model_path)
```

Step 1: Hand Gesture



```
1 list_label = label_dict_from_config_file("hand_gesture.yaml")
2 DATA_FOLDER_PATH=". ./data/"
3 testset = CustomImageDataset(os.path.join(DATA_FOLDER_PATH, "landmark_test.csv"))
4
5 # Test DataLoader instantiation
6 ##### Your Code Here ###### Q6
7 ''' Hoàn thành code bên dưới để khởi tạo DataLoader cho testset with batch size
8 20, không cho phép shuffle
9 '''
10 test_loader =
11 #####
```

Step 1: Hand Gesture



```
15 network = NeuralNetwork()
16 network.load_state_dict(torch.load(best_model_path, weights_only=False))
17
18 network.eval()
19 acc_test = Accuracy(num_classes=len(list_label), task='MULTICLASS')
20 for i, test_data in enumerate(test_loader):
21     test_input, test_label = test_data
22     ##### Your Code Here #####
23     '''Hoàn thành code bên dưới để predict class của cù'chỉ và update accuracy
24     với kết quả predict và true labels
25     '''
26     preds =
27
28     #####
29
30 print(network.__class__.__name__)
31 print(f"Accuracy of model:{acc_test.compute().item()}")
32 print("=====")
```

→ NeuralNetwork
Accuracy of model:0.9750000238418579
=====

Quiz

Model được sử dụng để phân loại cử chỉ có bao nhiêu hidden layer

- (A) . 2
- (B). 3
- (C). 4
- (D). 5

Step 2: Control Lights (Simulation + Reality)

➤ Simulation



```
step2_control_lights/
├── controller.py
└── detect_simulation.py
    └── hand_gesture.yaml
```

```
step2_control_lights/
├── controller.py
└── detect_simulation.py
    └── hand_gesture.yaml
        └── models
            └── model_03-11_04_37_NeuralNetwork_best
```

Best model từ step 1

Step 2: Control Lights (Simulation + Reality)

➤ Simulation



```
step2_control_lights/
└── controller.py
└── detect_simulation.py
└── hand_gesture.yaml
└── models
    └── model_03-11_04_37_NeuralNetwork_best
```

```
if __name__ == "__main__":
    model_path = "./models/model_02-11_15_19_NeuralNetwork_best"
    light = LightGesture(model_path, device=False)
    light.run()
```

Không chạy thiết bị
chỉ mô phỏng

Step 2: Control Lights (Simulation + Reality)

➤ Simulation



```
step2_control_lights/
└── controller.py
└── detect_simulation.py
└── hand_gesture.yaml
└── models
    └── model_03-11_04_37_NeuralNetwork_best
```

```
if __name__ == "__main__":
    model_path = "./models/model_02-11_15_19_NeuralNetwork_best"
    light = LightGesture(model_path, device=False)
    light.run()
```

Đường dẫn MLP model

Step 2: Control Lights (Simulation + Reality)

➤ Reality



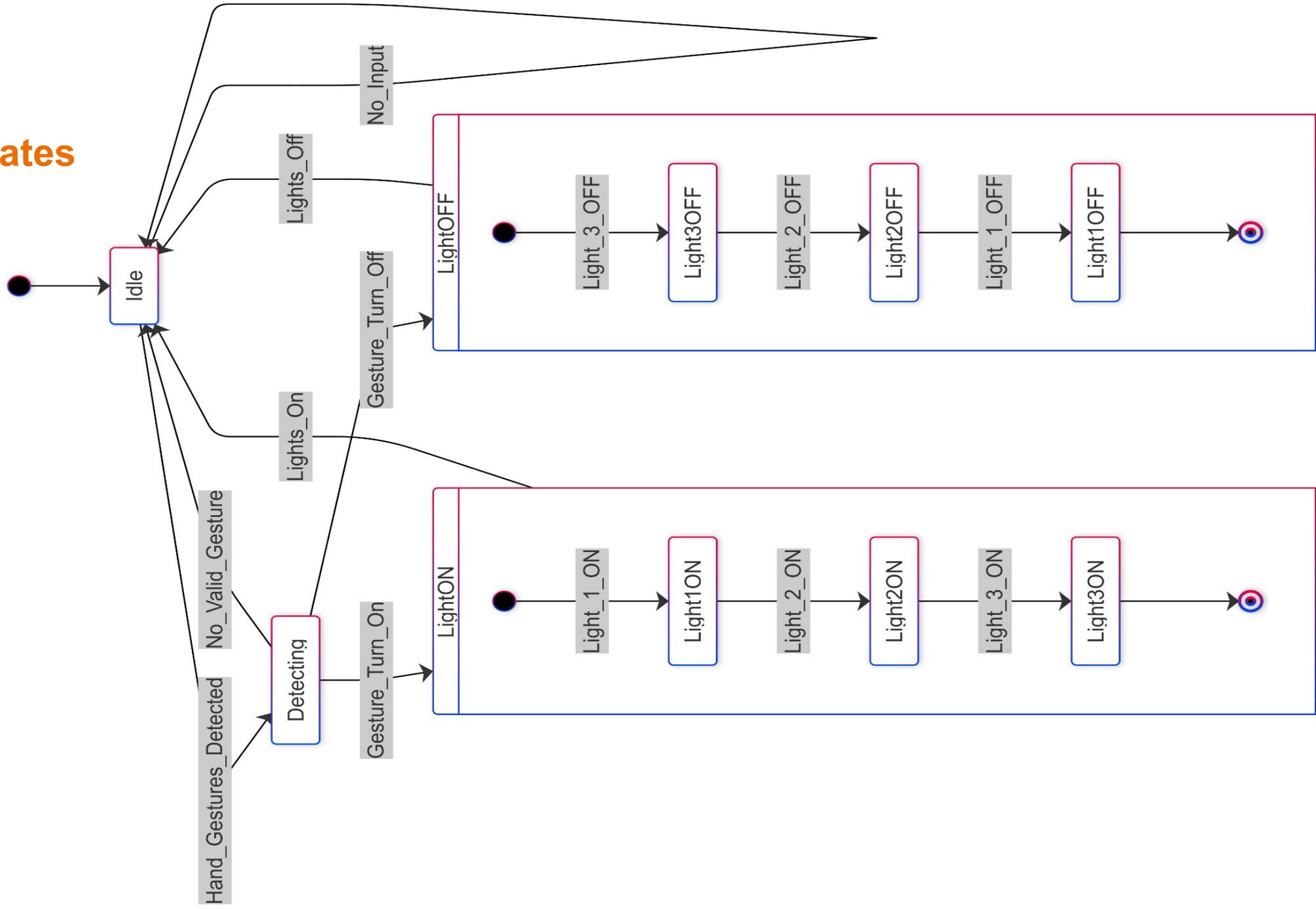
```
step2_control_lights/
└── controller.py
└── detect_simulation.py
└── hand_gesture.yaml
└── models
    └── model_03-11_04_37_NeuralNetwork_best
```

```
if __name__ == "__main__":
    model_path = "./models/model_02-11_15_19_NeuralNetwork_best"
    light = LightGesture(model_path, device=True)
    light.run()
```

Chạy với thiết bị



States

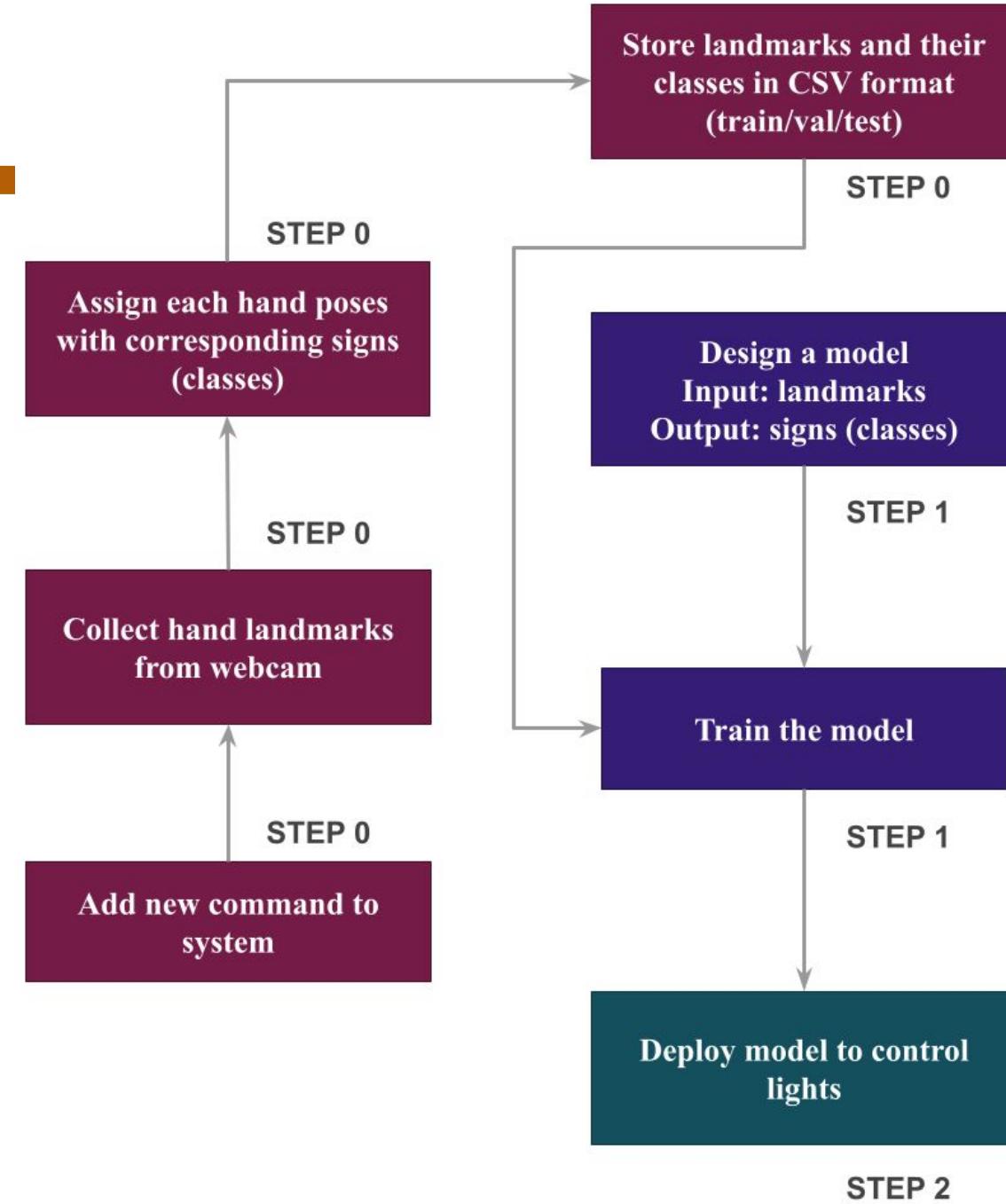


Quiz

Ở step 2 khi light 1 ở trạng thái ON và nhận lệnh bật light 2 thì trạng thái của light 1 sẽ thay đổi như thế nào

- (A) . Không thay đổi
- (B). Chuyển sang trạng thái ON
- (C). Chuyển sang trạng thái OFF
- (D). Không xác định

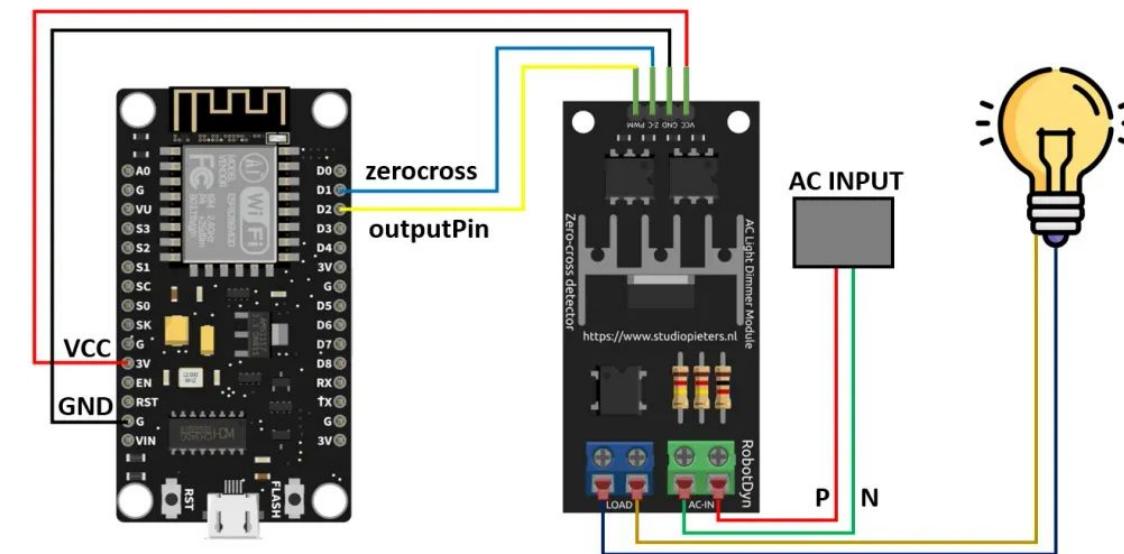
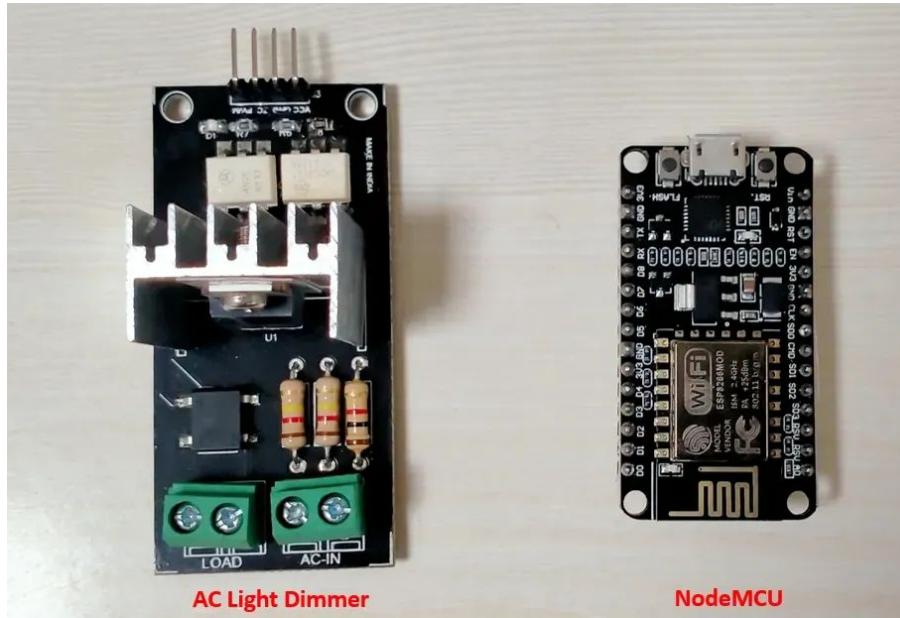
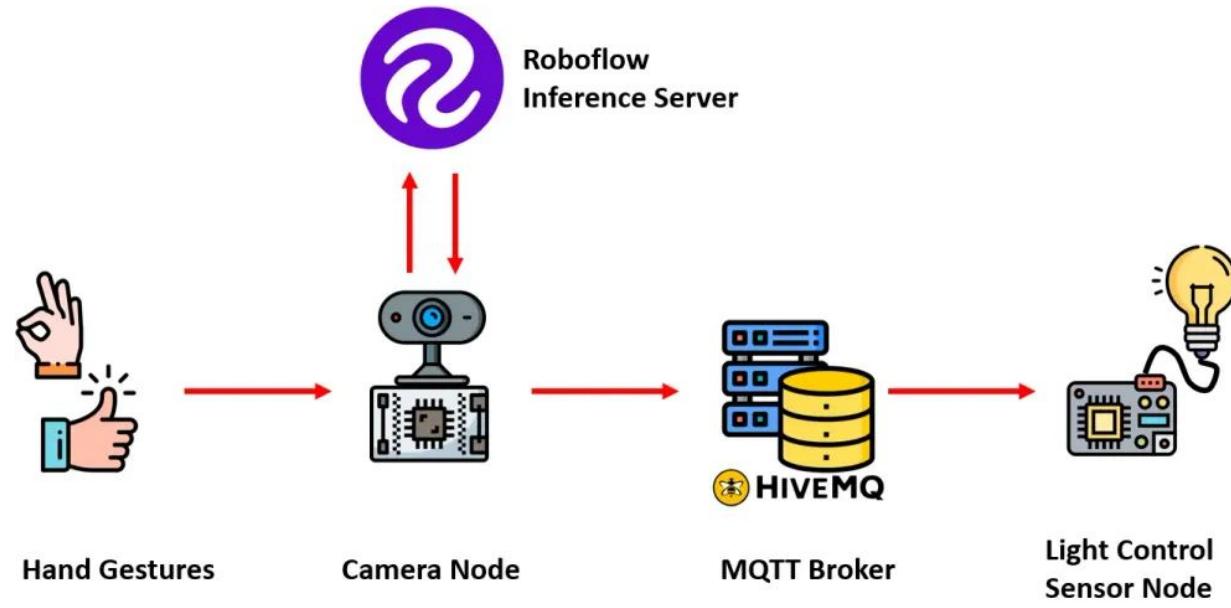
➤ Summary



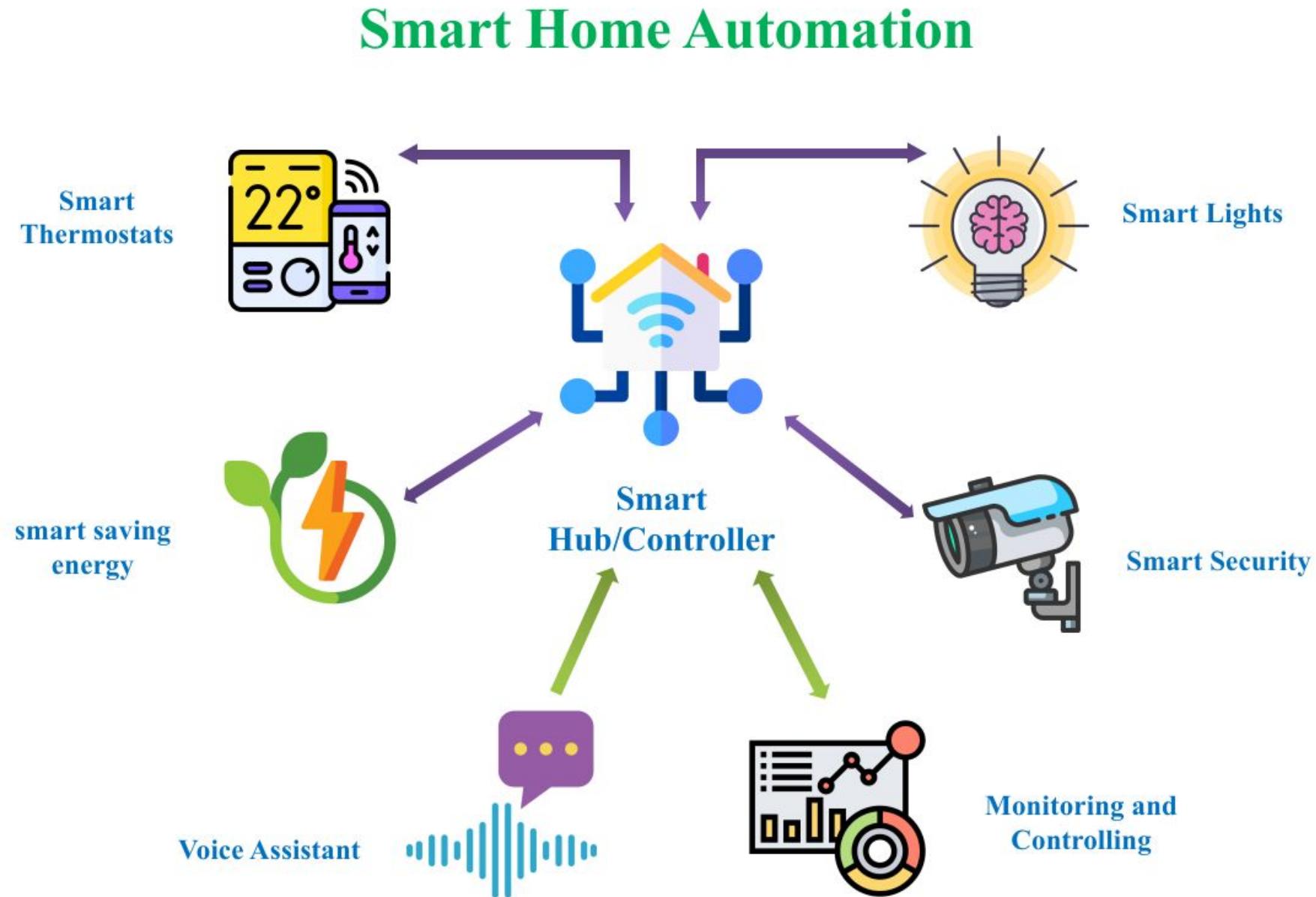
QUIZ



AI Application in Internet of Things (IoT)

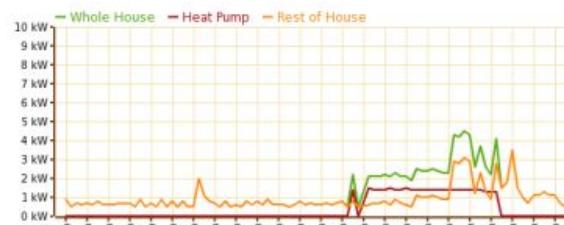
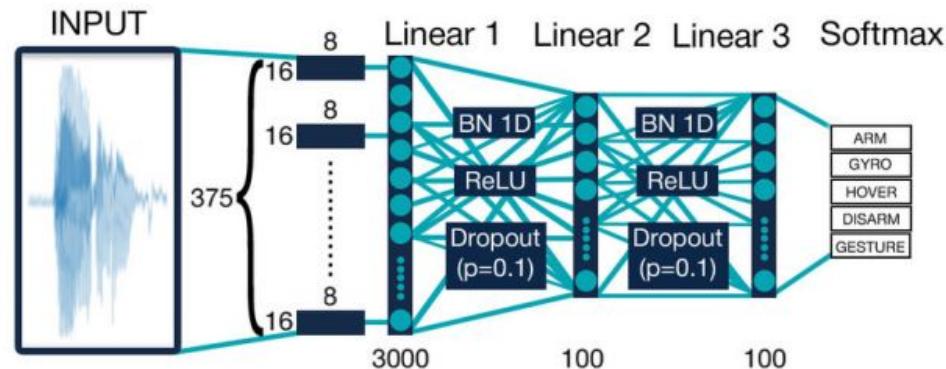


AI Application in Internet of Things (IoT)



Smart Home Automation Examples

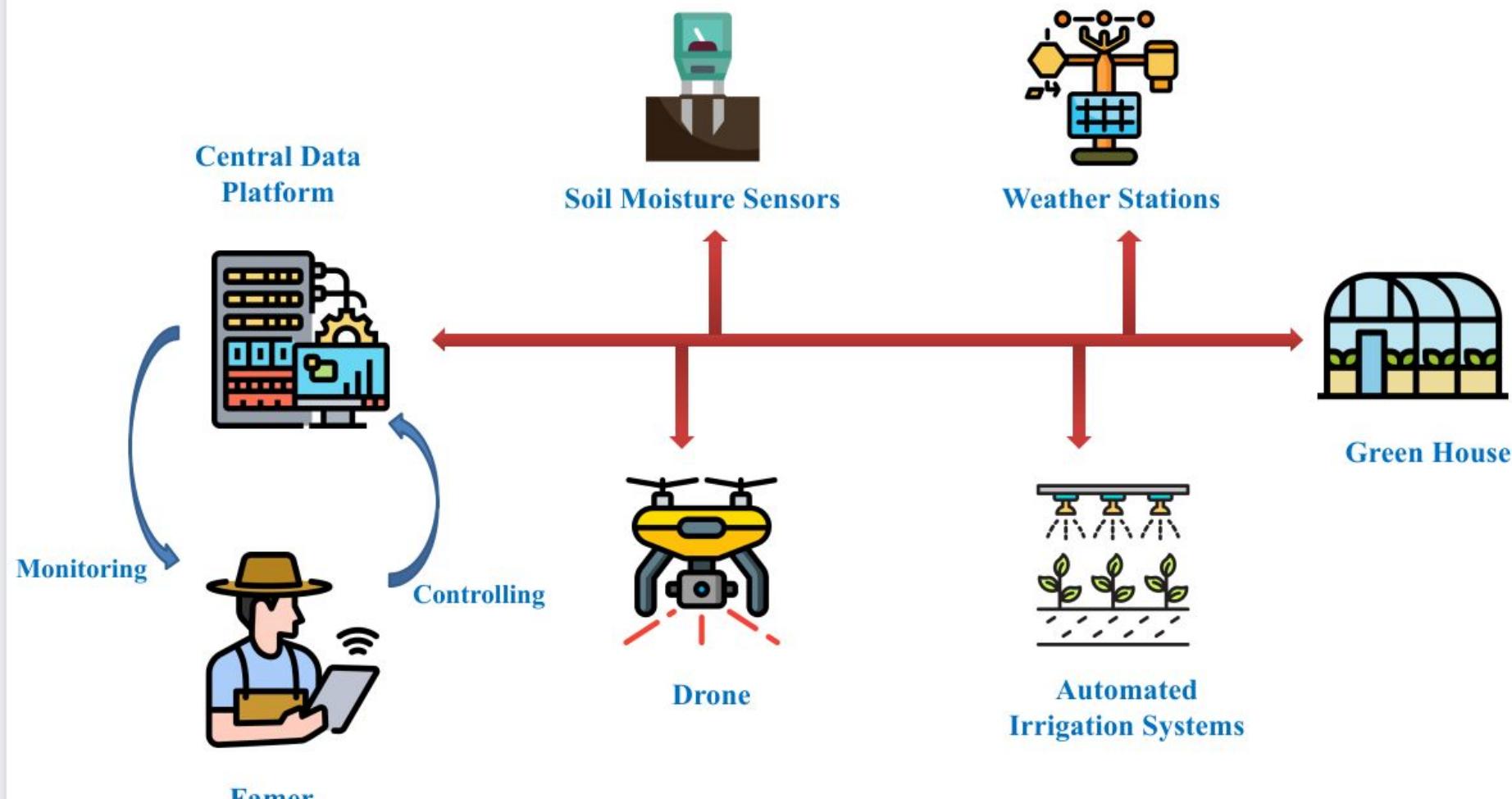
Voice Assistant



Monitoring & Controlling

AI Application in Internet of Things (IoT)

Smart Agriculture Ecosystem



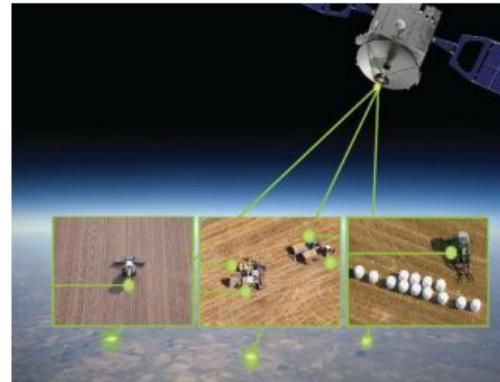
Smart Agriculture Ecosystem Examples



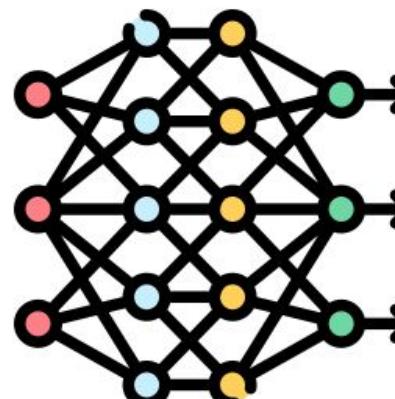
Drone Video Real-time



Object Detection



Datellite Images



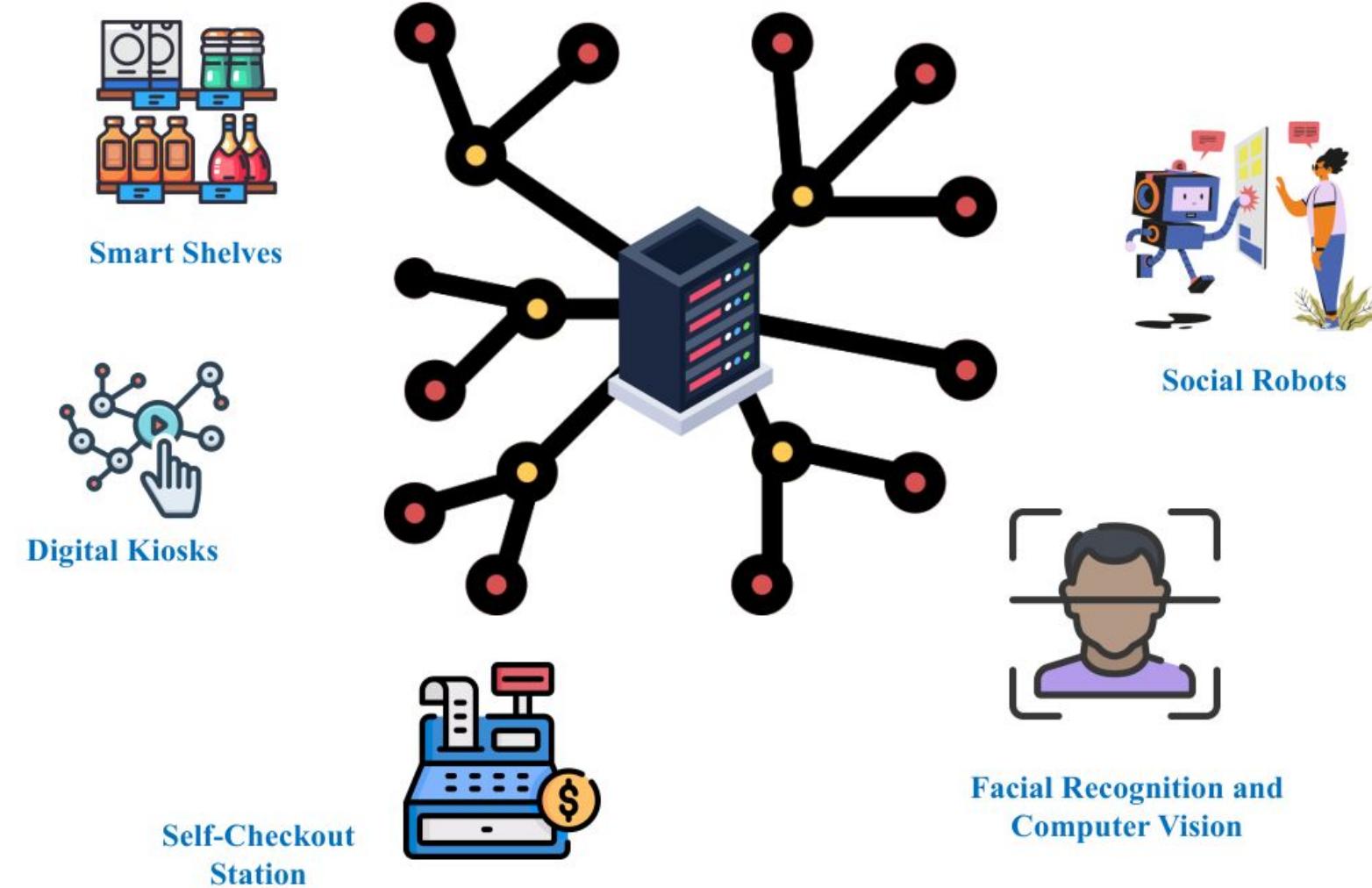
Deep Learning Model



Yield Estimation and Forecasting

AI Application in Internet of Things (IoT)

AI-Enhanced Retail Environment



AI Application in Internet of Things (IoT)

AI-Enhanced Retail Environment Examples



Self-Checkout Station



Facial Recognition and Computer Vision

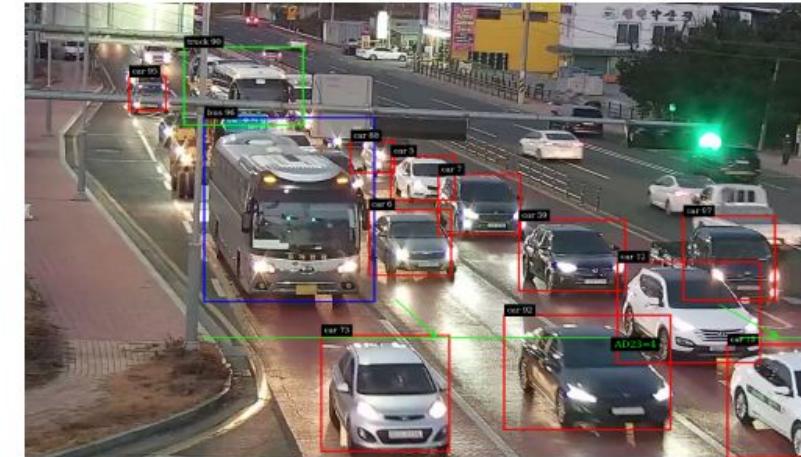
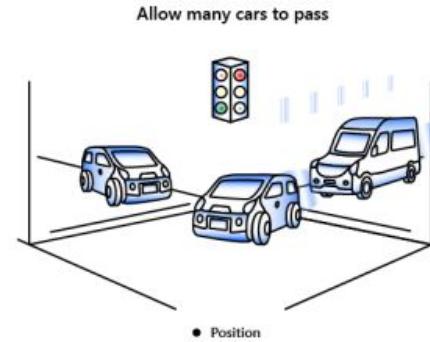
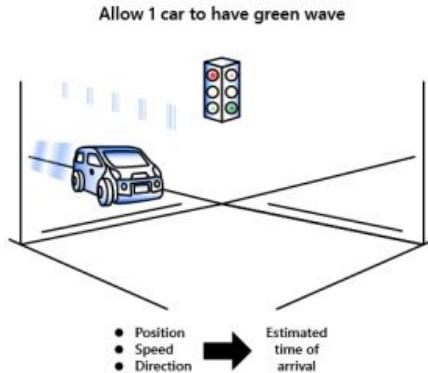


Smart Shelves

AI Application in Internet of Things (IoT)

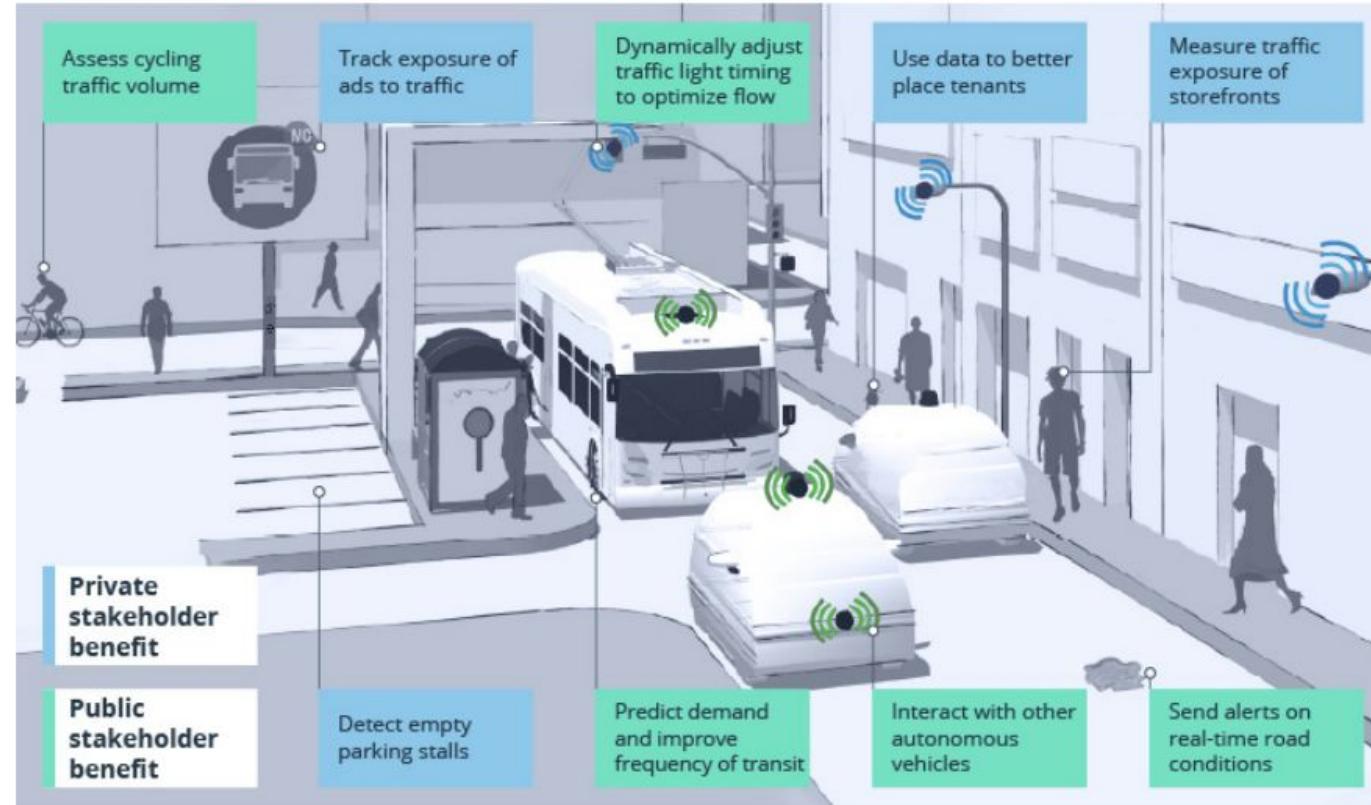


Smart City Examples



AI Application in Internet of Things (IoT)

Smart City Examples



AI-Driven Public Transportation
Optimization in Smart City

Objectives

- Develop a Natural Interaction System
- Set Up the Project Environment
- Collect and Prepare Gesture Data
- Build and Train an MLP Model
- Evaluate Model Performance
- Deploy Real-Time Gesture Recognition

