



# Enabling Ego Vision Applications on Smart Eyewear Devices

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## 1. Introduction to Ego Vision for Smart Eyewear

- What is Ego Vision
- Challenges of Embedded Vision
- Benefits of Ego Vision in Smart Eyewear
- Key Applications of Ego Vision

## 2. Ego Vision Functionalities: Ego Action, Human Pose Estimation, SLAM

## 3. Conclusions and Future Work

# Introduction to Ego Vision – What is Ego Vision?

11:08 AM  
Asakusa Temple

Hello, where is  
the station?

こんにちは、駅は  
どこですか。



# Introduction to Ego Vision – Benefits of Ego Vision in Smart Eyewear

## Hands-free functionality enhances usability

- Seamless user interaction
- Improved mobility
- Natural experience



## Real-time context-aware insights for decision-making

- Situational awareness
- Personalized information
- Actionable alerts



## Enhances accessibility and inclusivity

- Empowering the visually impaired
- Support for elderly users
- Breaking language barriers



## Hardware constraints

- Limited power and storage
- Small form factor
- Battery life constraints



SMART  
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## Real-time processing

- Real-time performance
- Balancing latency and computation



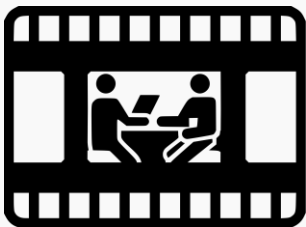
## Privacy and security

- Sensitive data
- Ensuring privacy
- Secure data transmission



## Scene understanding

- Interpret complex scenes
- Detect user context
- Generate audio-visual summaries



## Localization and mapping

- Real-time map of surroundings
- Outdoor and indoor navigation



## Activity recognition

- Understand user's actions
- Recognize hand gestures for interface control



# Ego Vision Functionalities – Human Pose Estimation

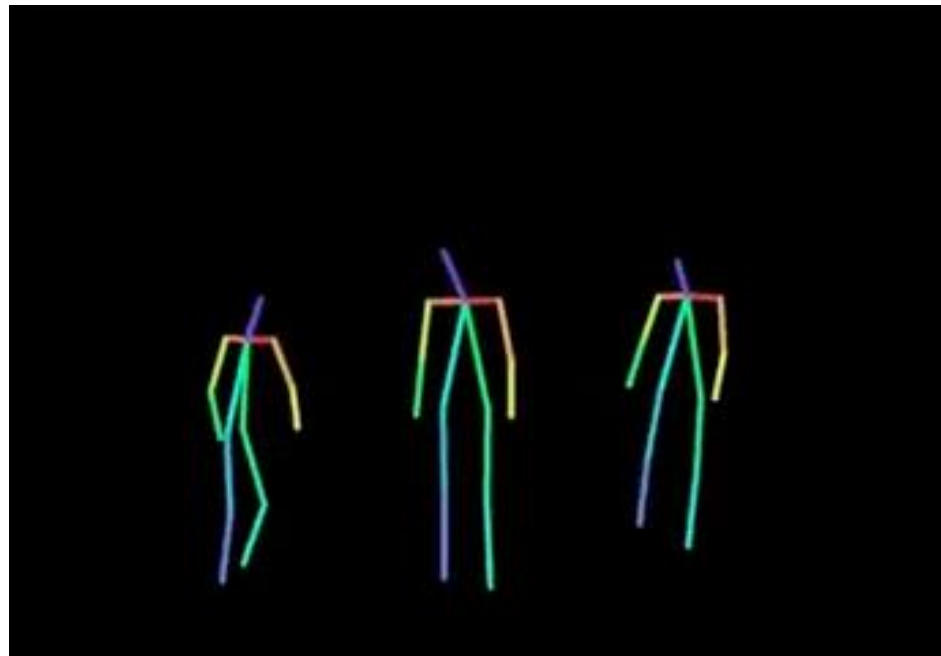
Detecting people's body movements to enhance user interaction and contextual awareness

## High-Precision Algorithms:

Multi-person pose detection optimized for mobile deployment [1, 2]

**Benchmarks Achieved:** 49% mAP on mobile, 60% mAP on edge

**Cons:** Missing real-time (30 FPS) on edge device (49 FPS on mobile, 11 FPS on edge)



<https://rf-action.csail.mit.edu/>

# Ego Vision Functionalities – Human Pose Estimation

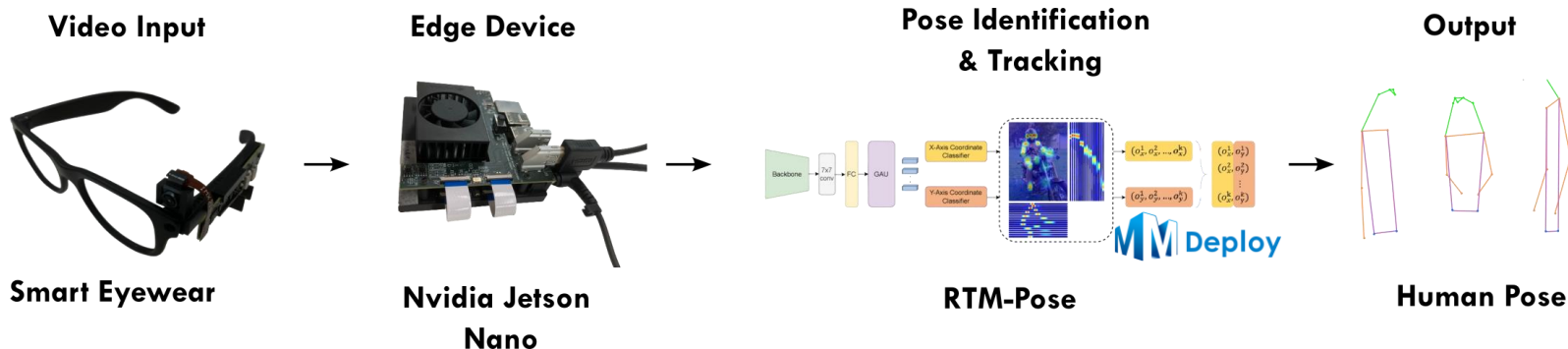
## Limitations

RTMPose on Nvidia Jetson Nano reaches 10 fps, insufficient for real-time processing

## Improvements

Converted RTMPose PyTorch model into C language with MMDeploy framework:

- Achieved over 30 fps on the Jetson Nano
- Process up to ten people simultaneously



H. Quan et al., Evaluating Human Pose Estimation Algorithms for Resource-Constrained Smart Eyewear Device, ECCV WS (2024)

# Ego Vision Functionalities – Human Pose Estimation



Model	<a href="#">Yolo Pose 8 (n)</a>	<a href="#">RTM Pose</a>	RTM Pose Opt C
mAP %	50.5	64.4	64.4
Hz	11	20	30

Developed model for Human Pose Estimation on smart eyewear achieving real-time speed (30 FPS)

**Future work:**  
Human Action Recognition

# Ego Vision Functionalities – Ego Action Recognition

Understanding and interpreting the user's actions from a first-person perspective

## Context-Aware Models:

First-person vision with multi-modal sensors [1, 2]

**Benchmarks Achieved:** >90% accuracy on large-scale ego-action datasets

**Cons:** Developed models not embeddable on edge

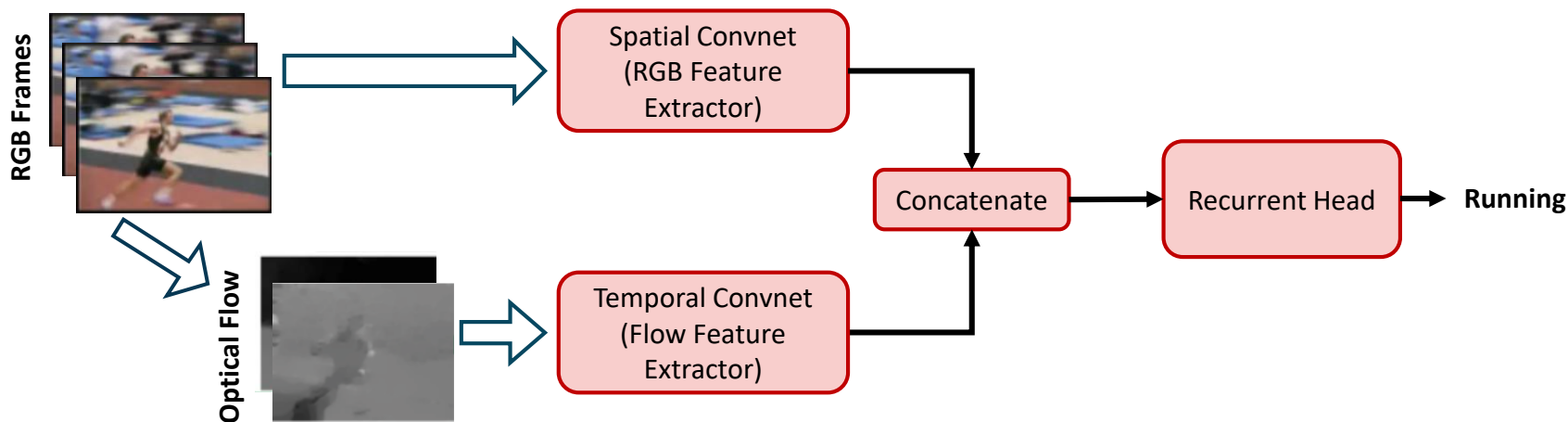


Damen et al., 2022

# Ego Vision Functionalities – Ego Action Recognition

**Goal: Understanding and interpreting the user's actions from a first-person perspective**

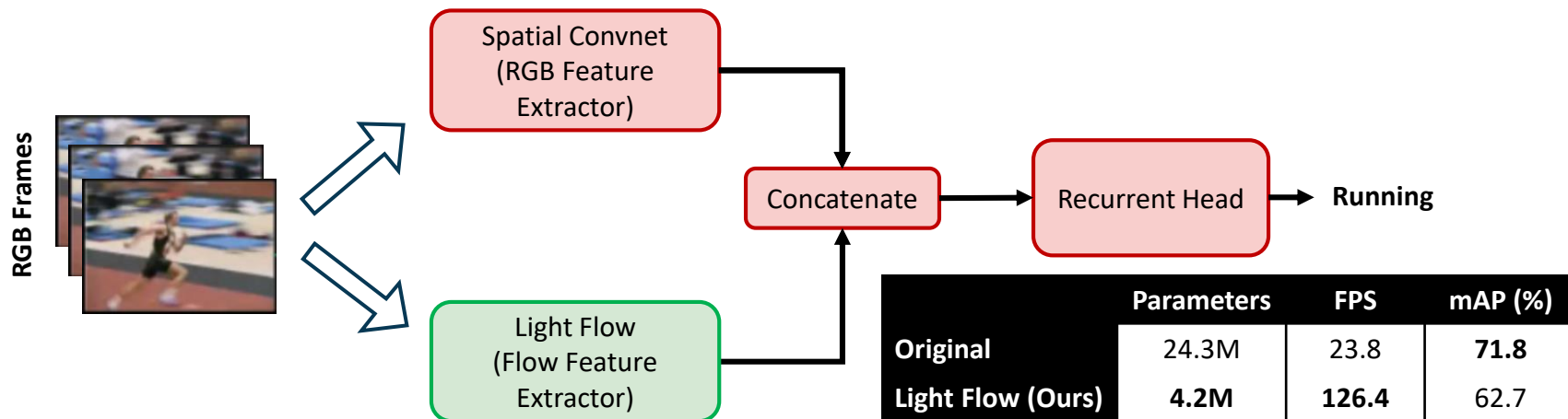
**MiniROAD** uses a two-stream Temporal Segment Network (TSN) to extract features from RGB and Optical Flow frames, followed by a recurrent head for action classification



# Ego Vision Functionalities – Ego Action Recognition

**Goal: Understanding and interpreting the user's actions from a first-person perspective**

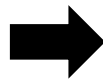
Optical Flow computation is costly, we replaced the temporal branch with a modified encoder from **Light Flow**, trained to extract optical flow features from RGB



# Ego Vision Functionalities – Ego Action Recognition

**Goal: Understanding and interpreting the user's actions from a first-person perspective**

- Jetson Orin Nano as computing edge device
- 3D printed eyewear frame + RGB camera



Model	<a href="#">LaViLa</a>	<a href="#">MiniROAD - L</a>
mAP %	-	76.2
Mean Acc	79.2	77.3
Hz	9.74	34.42



# Ego Vision Functionalities – SLAM

**Goal: Building a real-time map of the surroundings while tracking the user's position within it**

**Enhanced Localization and Mapping:** Meter-level mapping accuracy [1, 2]

**Benchmarks Achieved:** 13 FPS on edge

**Cons:** Models are difficult to embed and missing real-time (30 FPS) on edge device



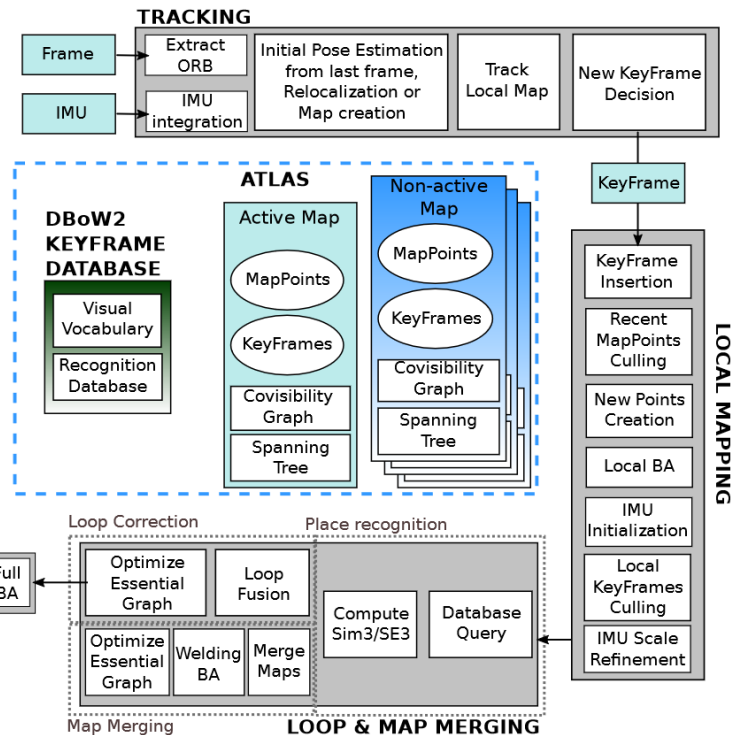
[SceneScript](#)

# Ego Vision Functionalities – SLAM

**Goal: Building a real-time map of the surroundings while tracking the user's position within it**

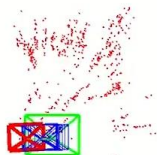
1. Able to work with different configurations
2. As fast as possible
3. Able to achieve high accuracy

**Identified Algorithm: Orb-SLAM 3**



# Ego Vision Functionalities – SLAM

We introduced a feedback loop to adaptively choose the number of points to track based on computation time



Model	<a href="#">ORB SLAM3</a>
Absolute Traj Error (m)	0.35 ± 0.58
Hz	34.42

SLAM MODE | Maps: 1, KFs: 10, MPs: 547, Matches: 198

ORB-SLAM3 algorithm for real-time indoor localization on edge devices (Jetson Orin Nano)

**Future work:** SLAM in dynamic environments

# Conclusions and Future Work

## Optimized Algorithms for Ego Action, Human Pose Estimation, and SLAM

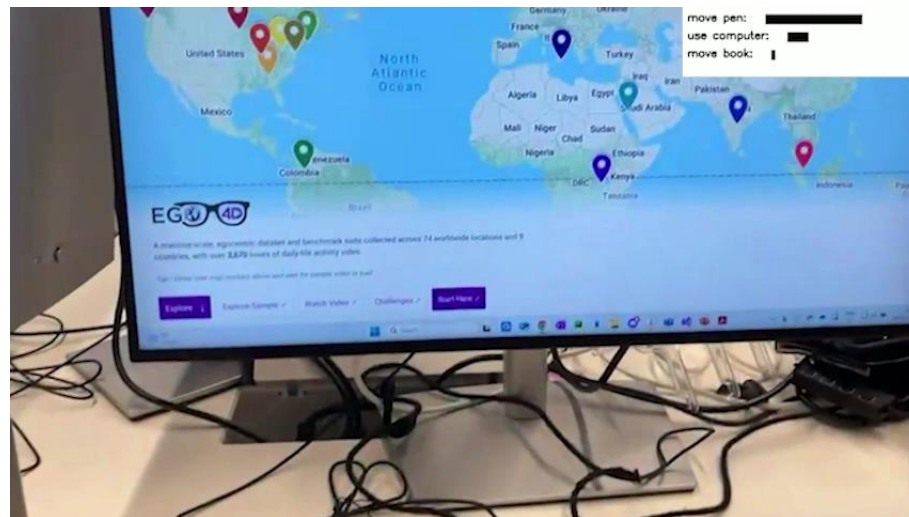
Achieved real-time performance (30 FPS) on Nvidia Jetson Orin Nano while balancing accuracy and efficiency

### Key Takeaways:

- Prioritize accuracy over generalization
- Optimization is essential
- Trade-offs must be carefully managed

### Future Work:

- Enhancing robustness (more scenarios)
- Expanding modalities (audio, IMU, etc.)



# Thank You for Your Attention!



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## Smart Eyewear Lab (Industry Research & Development)

- [Smart Eyewear Lab](#)

## Ego Action Recognition & Human Pose Estimation

- [Ego4D: Large-Scale Dataset for Egocentric Perception](#)

- [OpenPose: Multi-Person 2D Pose Estimation](#)

- [MMPose: Comprehensive Pose Estimation Framework](#)

## SLAM (Simultaneous Localization and Mapping)

- [ORB-SLAM3: State-of-the-Art Visual SLAM](#)

- [LIO-SAM: SLAM for LiDAR-IMU Fusion](#)