



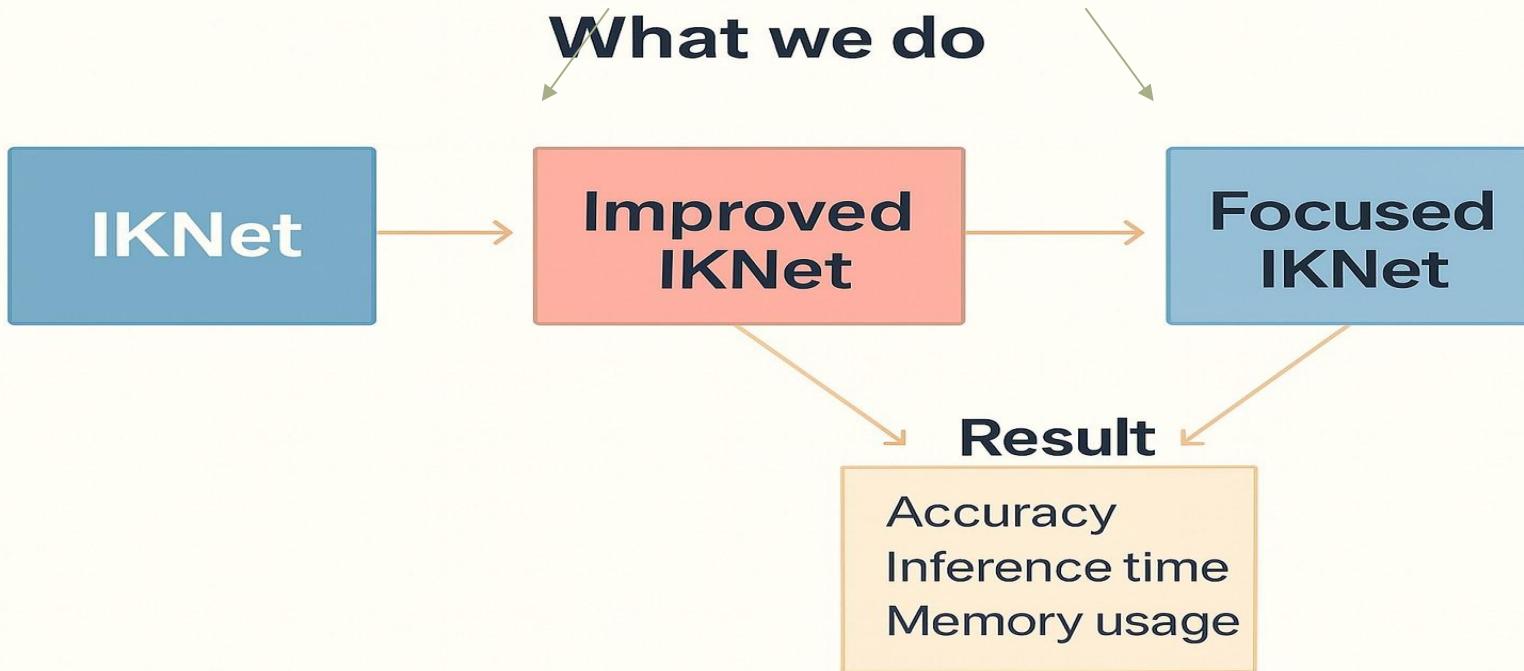
Inverse Kinematics Neural Network Models for Improving Inference Efficiency and Memory Usage

Sheng-Kai Chen | National Center for High-Performance Computing

Yi-Ling Tsai, Chun-Chih Chang | Electrical Engineering, Yuan Ze University

Yan-Chen Chen, Po-Chiang Lin

Outline



Background

Traditional IK Methods

- Computationally expensive and inefficient
- Unstable for complex systems

Deep Learning Models

- Faster and flexible
- Need a GPU

Motivation

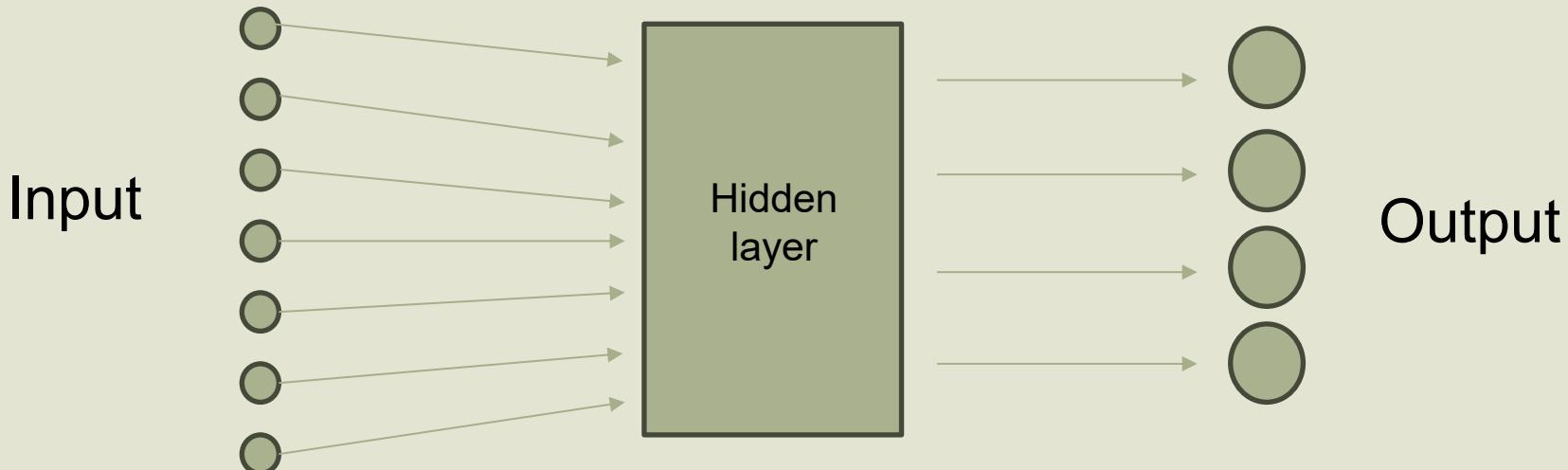
- CPU-only environments
- Reduce inference time and memory constraints

Original IKNet Overview

- Hurt performance
- Run slowly on the GPU and uses much memory

 [youtalk / iknet](#) Public

NVIDIA open source



Improved IKNet

Differences Between Original IKNet and Improved IKNet

- **Dropout only after the input layer**
- **Kaiming Normal initialization** improves weight scaling
- **BatchNorm1d** improves convergence and stability
- **Residual Blocks** to avoid the vanishing gradient problem

Focused IKNet

Differences Between Original IKNet and Focused IKNet

- Separate position and orientation
 - Each branch uses its layer
 - Combine output to MLP
- 
- Position 3D (x, y, z)
- Orientation 4D (qx, qy, qz, qw)

Experimental Setup

CPU

Intel Core i5-13500

GPU

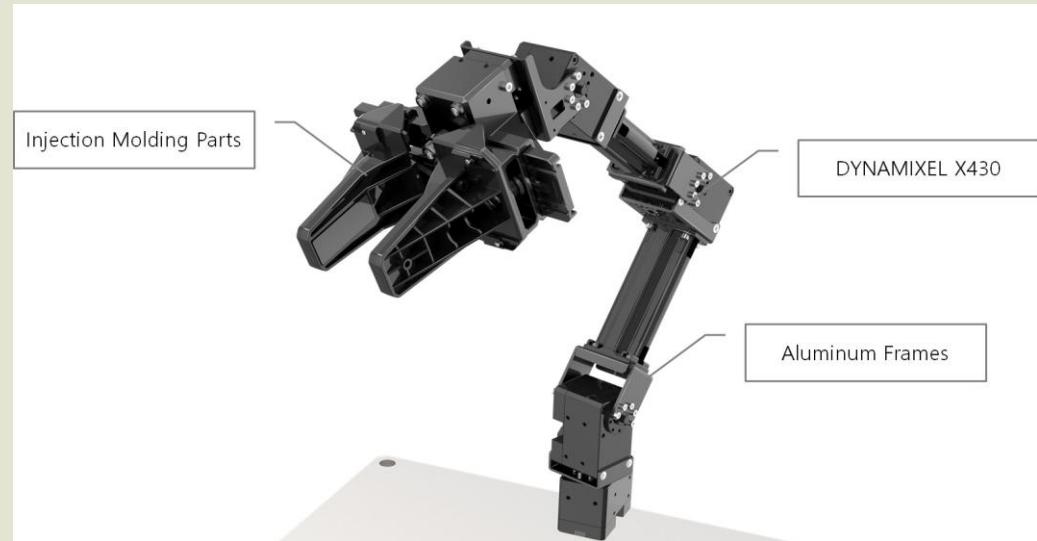
NVIDIA RTX 4060Ti

Memory

8 Gigabyte

Data collection

ROBOTIS Open Manipulator-X



Experimental Results & Analysis

Accuracy Comparison

Improved IKNet

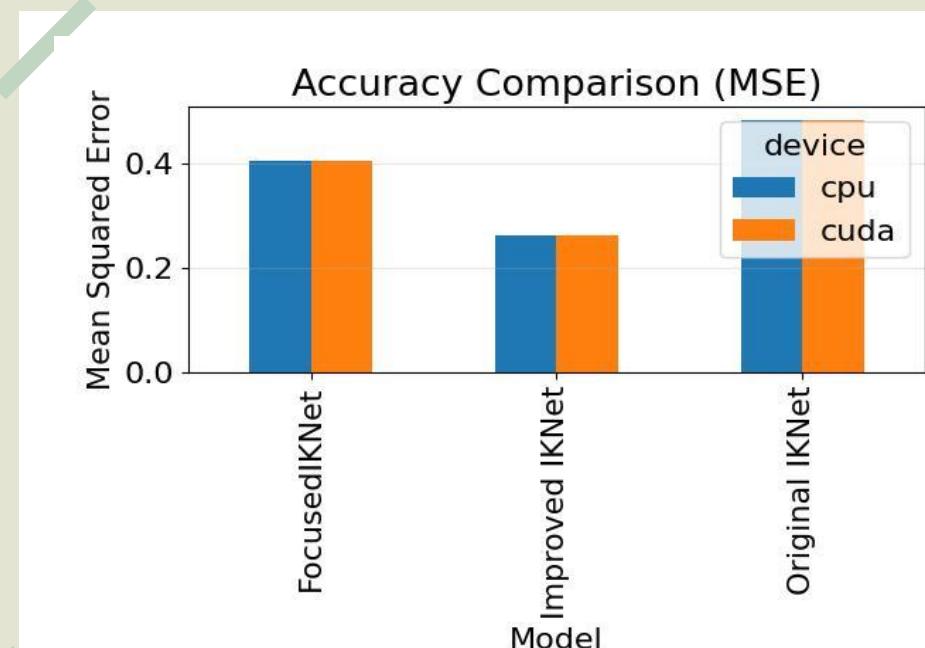
- Lowest MSE
- Predictive accuracy
- Model generalization

Focused IKNet

- Slightly higher MSE
- Dual-branch design offers flexibility
- Require further tuning

Original IKNet

- Worst performance on CPU



Inference Time Comparison

Improved IKNet

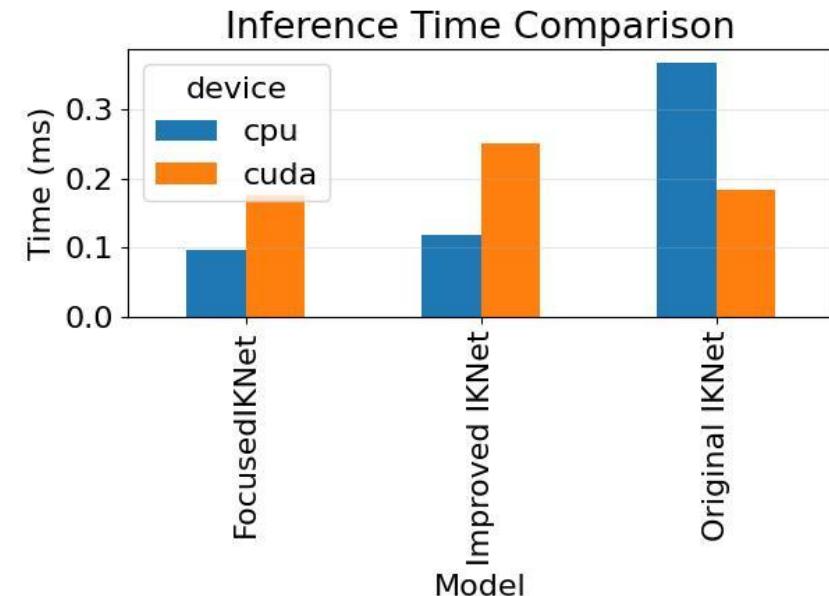
- Significantly faster on CPU
- Moderately faster on CUDA

Focused IKNet

- Shortest inference time
- Efficient execution speed

Original IKNet

- Slowest inference time on CPU
- Poor efficiency for real-time systems



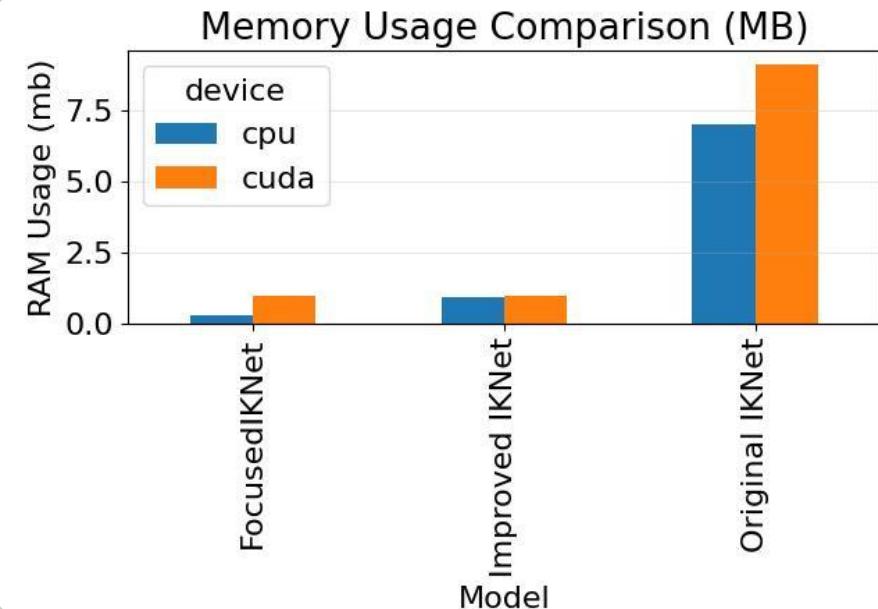
Memory Usage Comparison

Focused & Improved IKNet

- Less usage on both CPU and CUDA
- Highly optimized for efficient memory use
- Good accuracy and speed

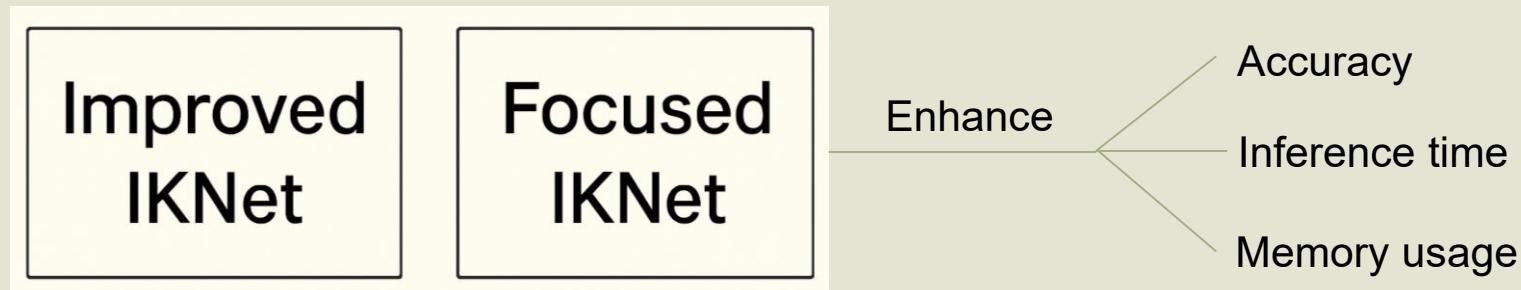
Original IKNet

- Larger model size
- Higher computational cost
- Unsuitable for memory-constrained embedded systems



Contributions

- CPU-only environments
- Enhance accuracy, inference time, and memory usage



CPU is better than GPU

Conclusion

Accuracy Comparison

- Most accurate model
- Well-suited for embedded CPU-only systems

Inference Time Comparison

- Ideal for low-latency applications (**Improved IKNet**)
- Offers a good trade-off between speed and accuracy (**Focus IKNet**)
- Not suitable for real-time CPU-based deployment (**Original IKNet**)

Memory Usage Comparison

- Low-memory environments (**Improved & Focused IKNet**)
- Resource-intensive (**Original IKNet**)

References

1.Deep-Neural-Network-for-Solving-the-Inverse-Kinematics Problem [Online] Available:

<https://github.com/OmarJItani/Deep Neural-Network-for-Solving-the-Inverse-Kinematics-Problem/blob/main/README.md>. (accessed in Feb 2025)

2.IKNet: Inverse kinematics neural networks [Online] Available:

<https://github.com/youtalk/IKNet?tab=readme-ov-file>. (accessed in Feb 2025)

3.CUDA[Online] Available:

<https://developer.nvidia.com/cuda-toolkit> (ac cessed in Feb 2025)

4.Pytorch [Online] Available:

<https://pytorch.org/> (accessed in Feb 2025)

Thank you for your listening!

Sheng-Kai Chen

2503100@niar.org.tw

Yi-Ling Tsai

yiling930910@gmail.com

Chen-Yan Chen

hannahchen0619@gmail.com

Chun-Chih Chang

Justin.chang93@gmail.com