



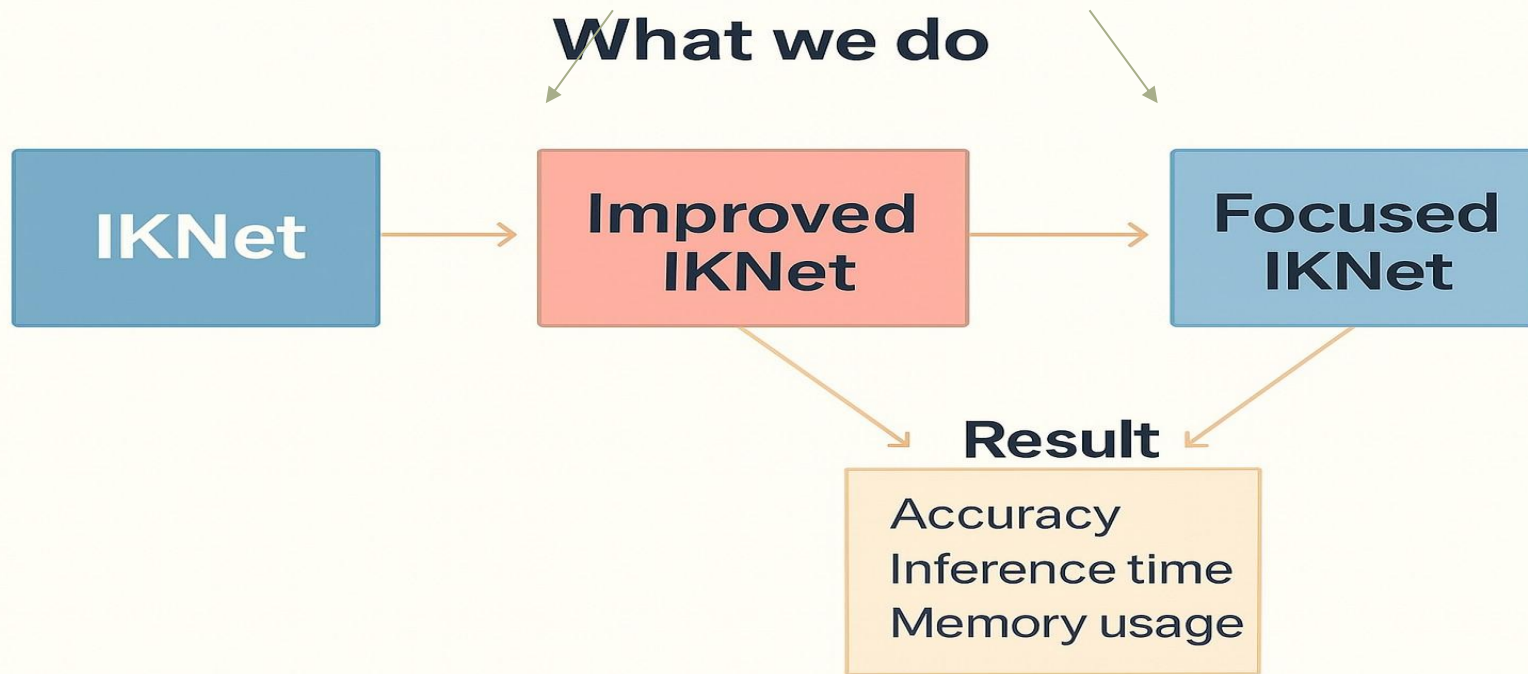
Inverse Kinematics Neural Network Models for Improving Inference Efficiency and Memory Usage

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Outline



Background

Traditional IK Methods

- Computationally expensive and inefficient
- Unstable for complex systems

Deep Learning Models


- Faster and flexible
- Need a GPU

Motivation

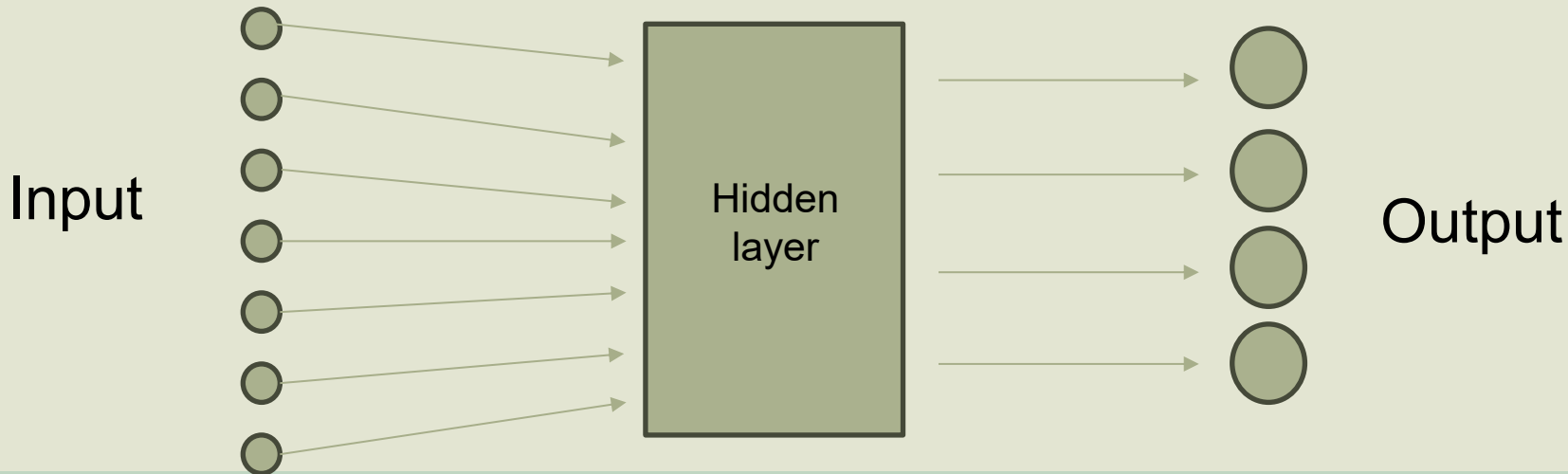
- CPU-only environments
- Reduce inference time and memory constraints

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

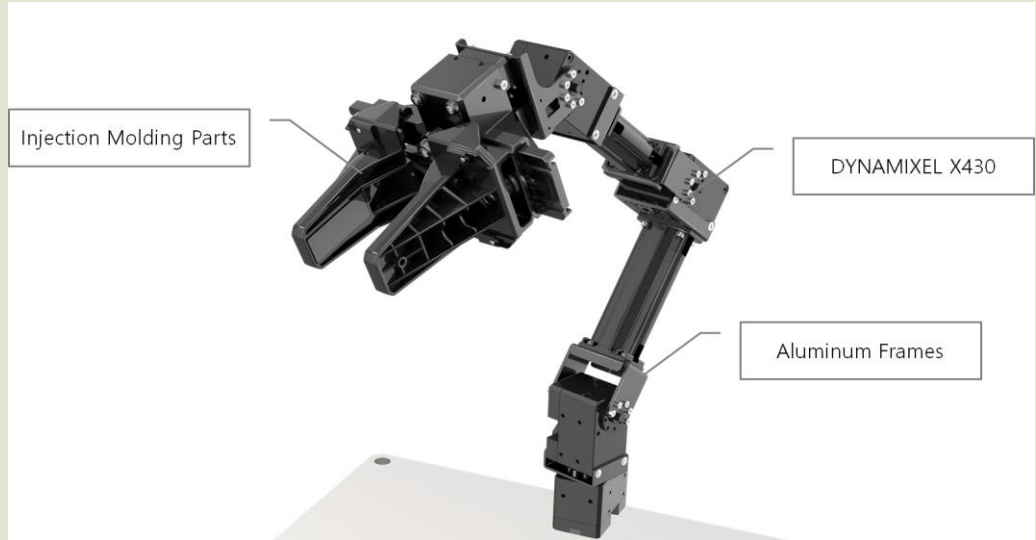
NVIDIA RTX 4060Ti


Memory

8 Gigabyte

Data collection

ROBOTIS Open Manipulator-X





Experimental Results & Analysis

Accuracy Comparison

Improved IKNet

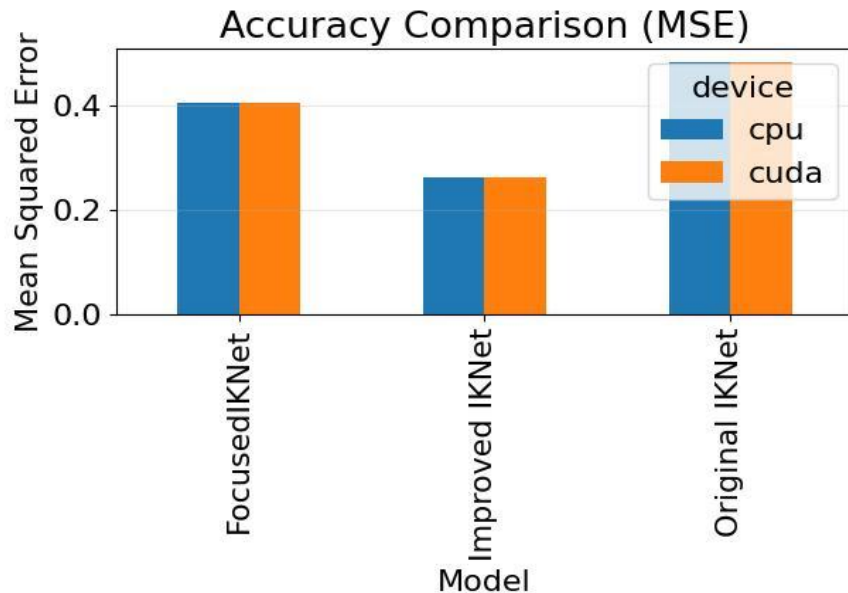
- Lowest MSE
- Predictive accuracy
- Model generalization

Focused IKNet

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

Original IKNet

- Worst performance on CPU



Inference Time Comparison

Improved IKNet

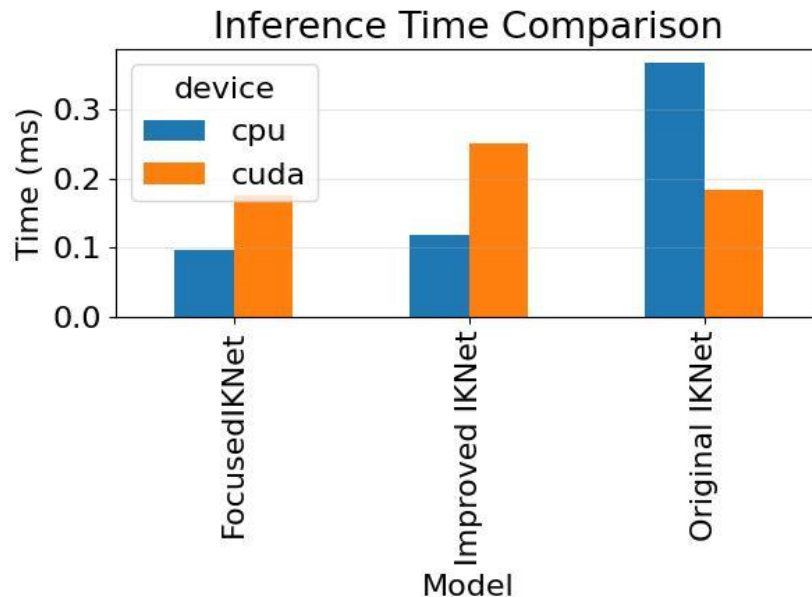
- Significantly faster on CPU
- Moderately faster on CUDA

Focused IKNet

- Shortest inference time
- Efficient execution speed

Original IKNet

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



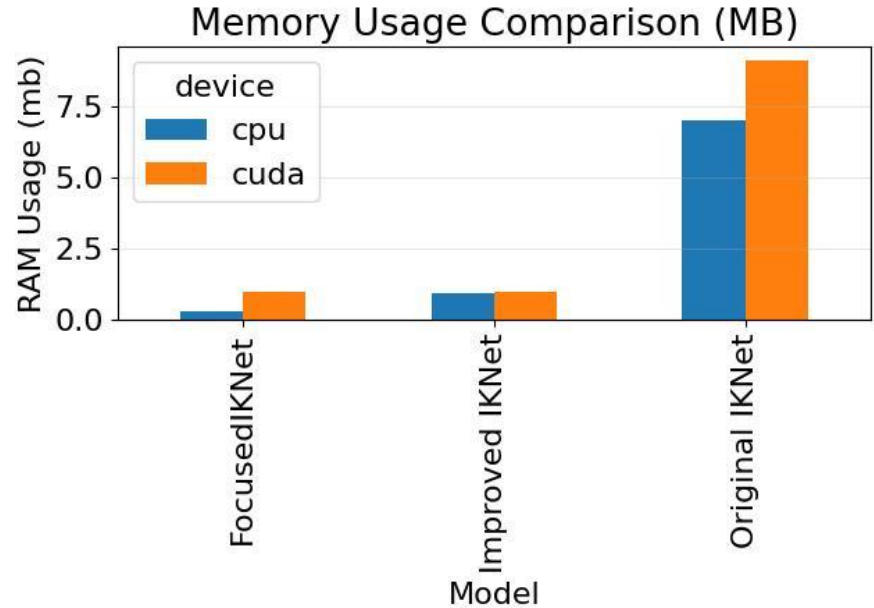
Memory Usage Comparison

Focused & Improved IKNet

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

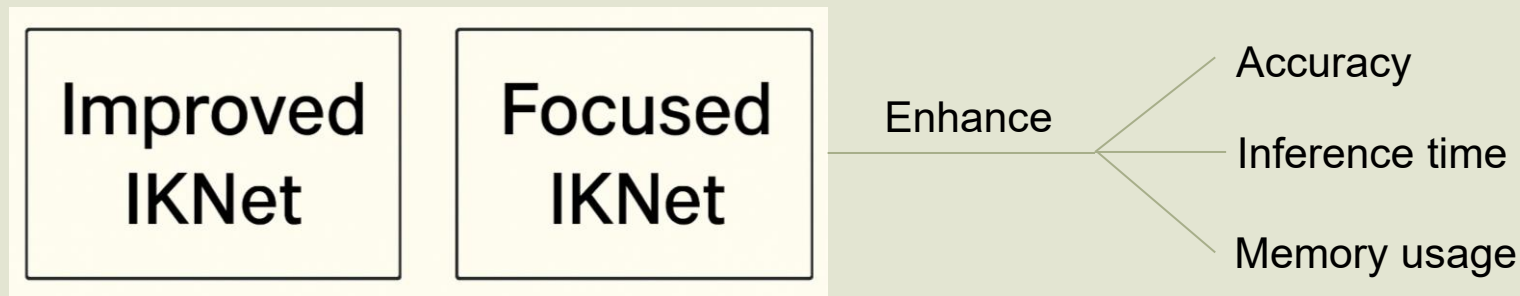
Original IKNet

- 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

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Thank you for your listening!

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