

A revolutionary approach to lightweight Al

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## Introduction

## Why Al Needs Optimization?

- $\rightarrow$  Al and CNNs drive innovation in industries like healthcare and autonomous driving.
- → CNNs are powerful but bulky, computationally expensive, and difficult to deploy on edge devices.

# The Problem

## Challenges of CNN Deployment

- → Bulky models with high computational demands.
- → Resource limitations on edge devices.
- → Need for efficient, lightweight alternatives without losing accuracy.

# The Solution

#### **Enter RL-Pruner**

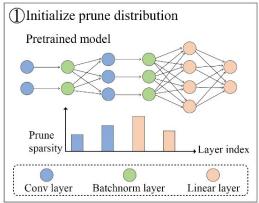
- → Uses reinforcement learning to dynamically prune unnecessary filters.
- → Creates smaller, faster CNNs with minimal accuracy loss.
- → Efficient for real-world applications like IoT, wearables, and autonomous vehicles.

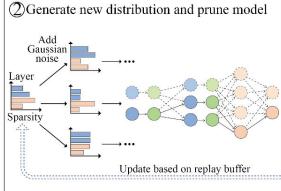
## **How RL-Pruner Works**

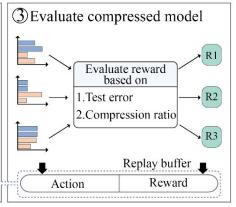
### The Four-Step Process

- → Dependency Graph: Maps layer relationships to preserve critical connections.
- $\rightarrow$  Layer-Wise Sparsity: Dynamically adjusts pruning per layer using reinforcement learning.
- → Layer Pruning: Removes less critical filters using the Taylor criterion.
- ightarrow Knowledge Distillation: Fine-tunes the pruned model with guidance from the original model.

# **Working Diagram**







# Results That Matter

### RL-Pruner's Key Achievements

- → High Compression with Minimal Performance Loss:
  - VGG-19: 60% sparsity with < 1% accuracy drop.</li>
  - MobileNetV3-Large: 40% sparsity with minimal loss.
- → Superior to Other Methods:
  - Outperforms DepGraph and GNN-RL on accuracy and compression ratios.

# Real-World Applications

### Transforming Industries with RL-Pruner

- → Faster Image Recognition: Real-time processing for drones and cameras.
- → Wearables: Efficient AI for health monitoring and sensing.
- → Autonomous Vehicles: Reduces latency and ensures safer decision-making.







# Conclusion

### The Future of Lightweight Al

- $\rightarrow$  RL-Pruner combines reinforcement learning and structured pruning for efficient Al.
- → Enables powerful models on resource-constrained devices.
- → Opens new possibilities for sustainable and accessible AI solutions.

