IOS: Inter-Operator Scheduler for CNN Acceleration

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Executive Summary

Motivation



Inter-Operator Scheduler



1.1-1.5x speedup

Efficient Deployment of CNNs is Important



Face Recognition

Self Driving

Language Translation

Is CNN inference in current DL libraries well utilizing underlying hardware?

Motivation for Inter-Operator Parallelization

1. More small convs in CNN design.



2. GPU peak performance increased



3. Intra- and Inter-operator Parallelization



Background: Wavefront Schedule Policy

Wavefront Schedule Policy: Execute all available operators stage by stage



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Wavefront schedule policy is sub-optimal

General Idea: Explore the schedule space exhaustively.



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Challenge: The number of schedules is exponential in the number of operators.

e.g., NASNet has more than 10¹² schedules
Prohibitive to enumerate

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Observation 2: The width of the computation graph is usually small



Maximum number of parallelizable operators

The width of Inception V3 is 6.

Inception Block

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Observation 1: Optimal schedule for a subgraph can be reused **Key Idea**: Dynamic Programming

Observation 2: The width of the computation graph is usually small **Key Result**: Time complexity is only exponential in the width



Parallelization Strategy Selection

 $Latency[S] = \min_{\substack{S' \text{ can be a last stage of } S}} (Latency[S-S'] + \frac{StageLatency(S')}{S' \operatorname{can be a last stage of } S})$



Last Stage Candidates

 $Latency[S] = \min_{\substack{S' \text{ can be a last stage of } S}} (Latency[S - S'] + StageLatency(S'))$

S' can be a last stage of $S \iff$ There is no edge from S' to S - S'

Edge from S - S' to S'

Edge from S' to S - S'



a b a b c d c d s' f g e f g



Operators *S* to be scheduled

S' can be a last stage of S

S'' can **NOT** be a last stage of S

Transition Graph and Time Complexity



Transition Graph and Time Complexity



Transition Graph

Transition Graph and Time Complexity



Methodology

Benchmarks •

•

- Inception V3 Fxpert Designed ۲
- SqueezeNet •
- ۲
 - Randwire Neural Architecture Search NasNet

Baselines

State-of-the-art Frameworks (cuDNN-based)



Different schedules on IOS Runtime ٠

Environment



- **IOS Implementation** •
 - cuDNN kernels
 - **CUDA Stream** •

Comparison of cuDNN-based Frameworks

Tensorflow: A popular machine learning framework.
 Tensorflow-XLA: TensorFlow with compilation optimization.
 TASO: Transformation-based optimizer.
 TVM-cuDNN: TVM backed with cuDNN convolution kernel.
 TensorRT: NVIDIA high-performance inference engine.
 IOS: Our method





IOS outperforms all frameworks and achieves **1.1-1.5x** speedup.

Performance is normalized to the best framework

Comparison of Different Schedules



Performance is normalized to the best schedule

More Active Warps Improve Utilization



Conclusion

- Sequential execution suffers from under utilization problem.
- Inter-Operator Scheduler (IOS):
 - Utilize both intra- and **inter-operator parallelism** in CNNs.
 - Dynamic-programming explores the schedule space exhaustively.
 - Time Complexity: $O\left((n/d+1)^{2d}\right)$, d is usually small.
- Key Results: 1.1-1.5x speedup on diverse CNNs.



