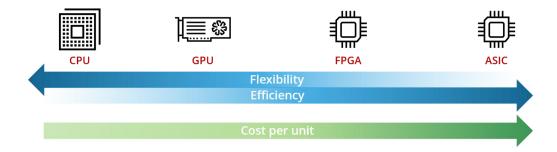
TVM: End to End Deep Learning Compiler Stack

Alina Egorova Arathy Ajaya Kumar

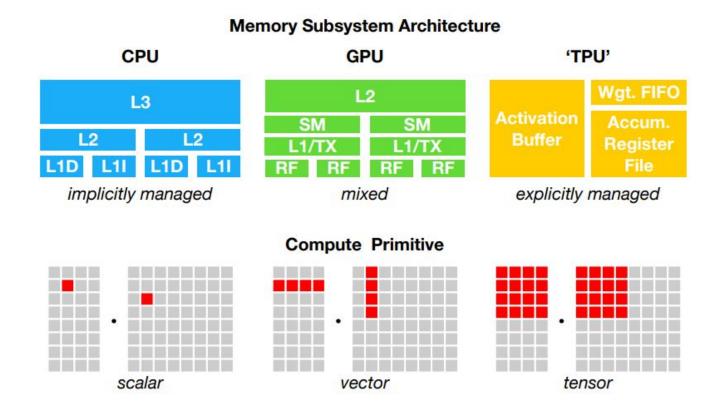
The challenges that deep learning is facing today







The challenges that deep learning is facing today



TVM: End to End Optimization Stack

Frameworks





Pytorch, caffe2, cntk supported via onnx



CoreML





Computational Graph Optimization

Tensor Expression Language

Primitives in prior works Halide, Loopy

New primitives for GPU Accelerators Loop Transformations Thread Bindings

Cache Locality

Thread Cooperation

Tensorization

Latency Hiding













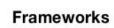








TVM: End to End Optimization Stack







CNTK



CoreML





Computational Graph Optimization

- operator fusion
- constant-folding
- static memory planning pass
- data layout transformations











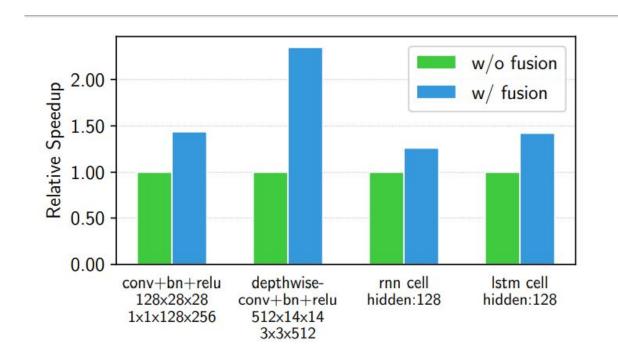






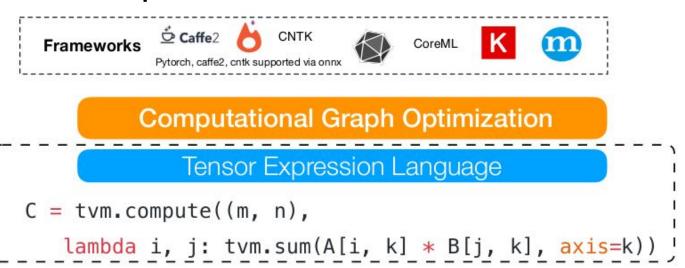


Operator fusion



Performance comparison between fused and non-fused operations. TVM generates both operations. Tested on NVIDIA Titan X

TVM: End to End Optimization Stack



Schedule Optimizations

Hardware

















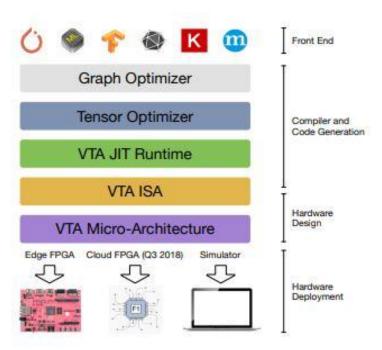
Schedule Optimizations

Schedule primitives used in various hardware backends	CPU Schedule	GPU Schedule	Accel. Schedule
[Halide] Loop Transformations	V	√	V
[Halide] Thread Binding	V	V	√
[Halide] Compute Locality	√	V	√
[TVM] Special Memory Scope		V	√
[TVM] Tensorization	V	V	√
[TVM] Latency Hiding			√

Versatile tensor accelerator (VTA)

- An extension of TVM stack
- Exposes salient features of deep learning accelerators.
- Provides an open deep learning system stack for optimizations.

VTA stack overview



VTA stack overview

- 1. NNVM (Neural network virtual machine)Intermediate Representation
- 2. TVM Intermediate Representation
- 3. VTA JIT Runtime
- 4. VTA Instruction Set Architecture
 - high-level CISC ISA
 - a low-level, and fixed latency RISC ISA
- 5. VTA Hardware Micro-Architecture