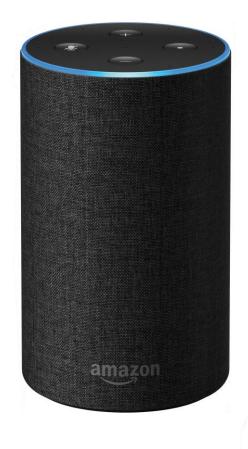


FW NXT

Snowflake deep neural network accelerator



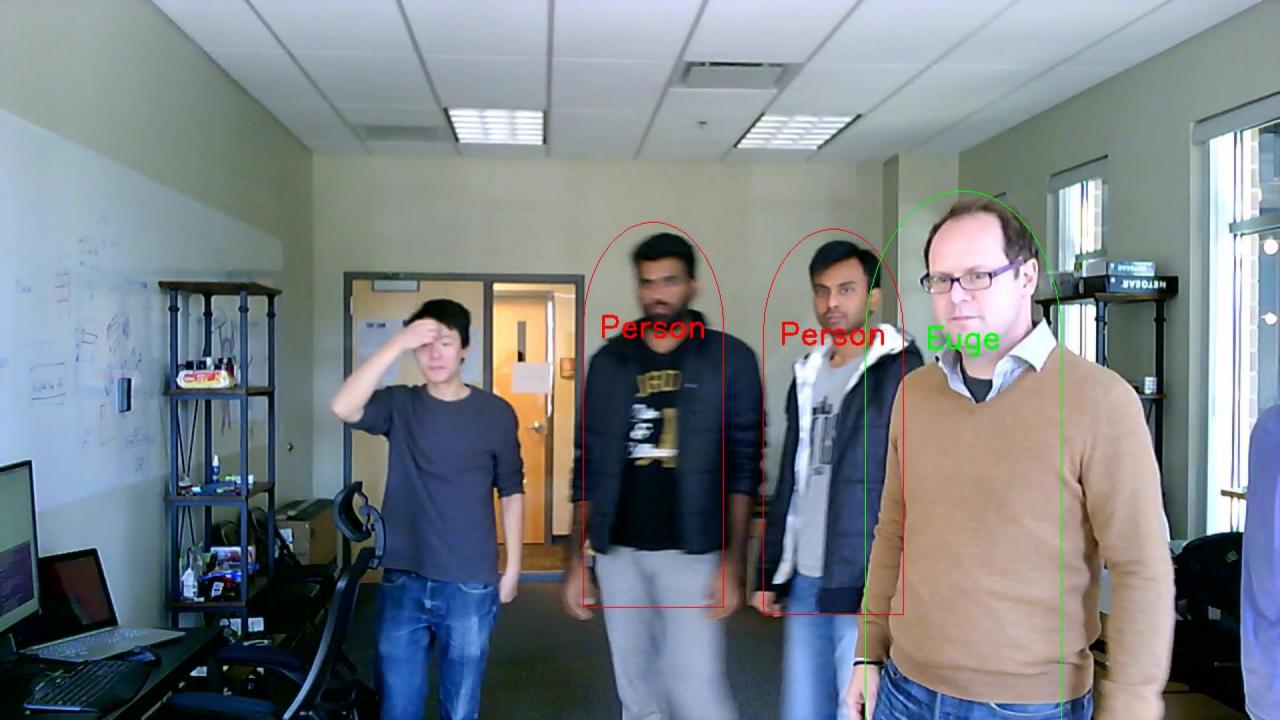
Deep Learning

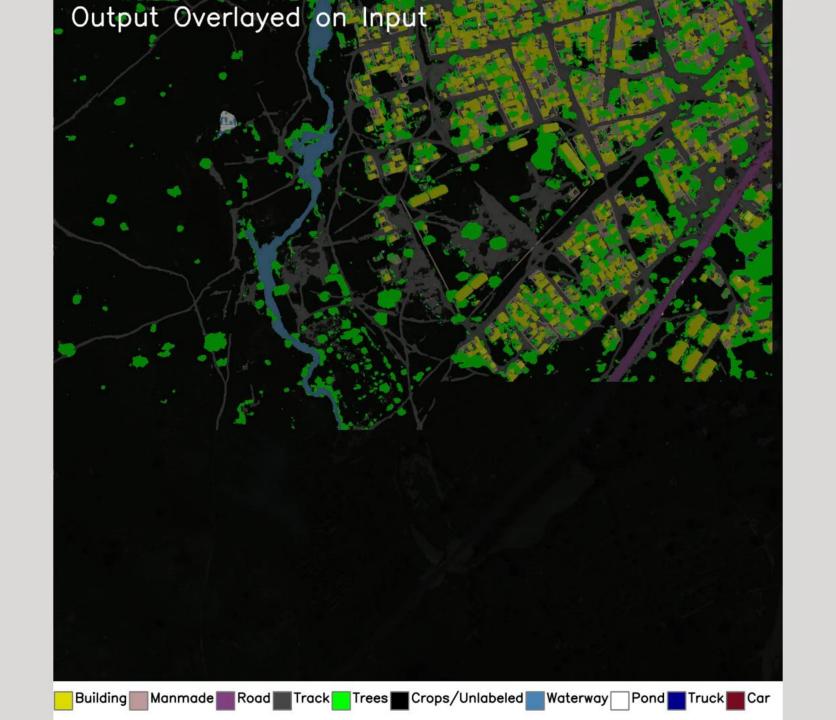






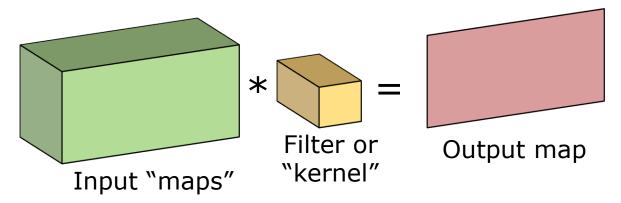




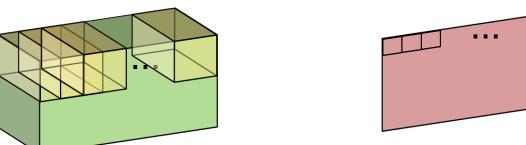


Input Image (3*224*224)Conv (11x11 s4) Maxpool (3x3 s2) Conv (5x5 s1) Maxpool (3x3 s2) Conv (3x3 s1) Conv (3x3 s1) Conv (3x3 s1) FCN (4096) FCN (4096) FC (1000) Prediction

Convolutional Neural Networks

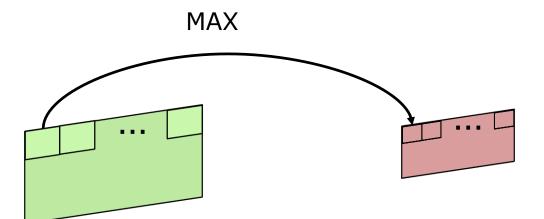


- Compute intensive
- Embarrassingly parallel
- Comprise > 95% of the workload
- Comprised of mult-acc (MAC) ops



Input Image (3*224*224) Conv (11x11 s4) Maxpool (3x3 s2) Conv (5x5 s1) Maxpool (3x3 s2) Conv (3x3 s1) Conv (3x3 s1) Conv (3x3 s1) FCN (4096) FCN (4096) FC (1000) Prediction

Convolutional Neural Networks



- Make up ~1% of the workload
- Lesser parallelism to exploit
- Comprised of comparisons

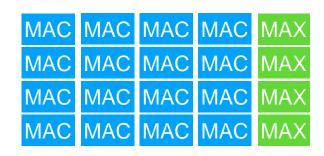
Input Image (3*224*224) Conv (11x11 s4) Maxpool (3x3 s2) Conv (5x5 s1) Maxpool (3x3 s2) Conv (3x3 s1) Conv (3x3 s1) Conv (3x3 s1) FCN (4096) FCN (4096) FC (1000) Prediction

Convolutional Neural Networks



- Tens of MB of weights
- No weights reuse
- Bandwidth intensive
- Comprised of MACs

Accelerator Hardware









Functional units

- Multiply-accumulate (MAC)
- Comparators (maxpool)

On-chip memory

Buffers for maps and weights

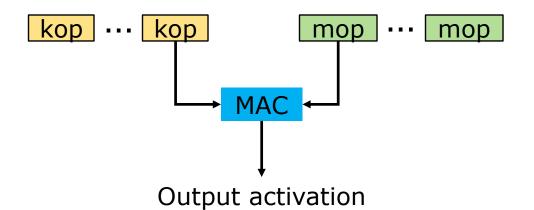
Configuration logic

- Instruct on-chip memory to stream to MACs
- Instruct MACs to write-back results

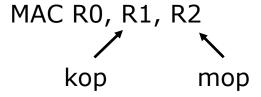
Types of Parallelism

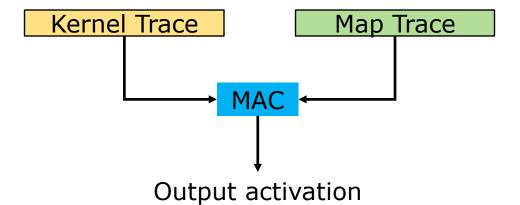
Intra-map, intra-activation MACs share both input operands MACs share weights Intra-map, inter-activation Inter-map No data sharing Inter-layer No data sharing © FWDNXT 2018

Traces

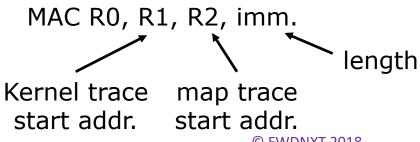


Require N instrs 1 instr per cycle



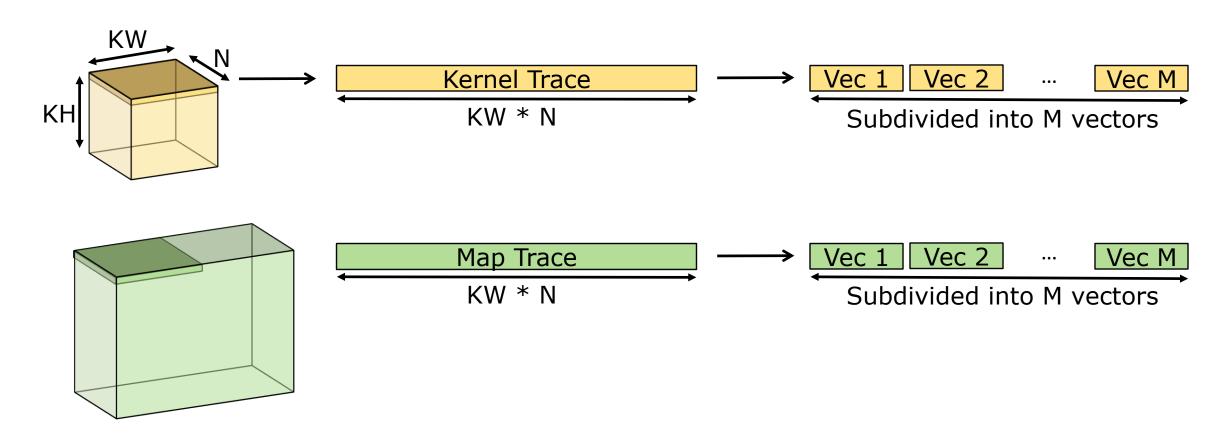


Require 1 instr 1 instr per trace Require start addr, length

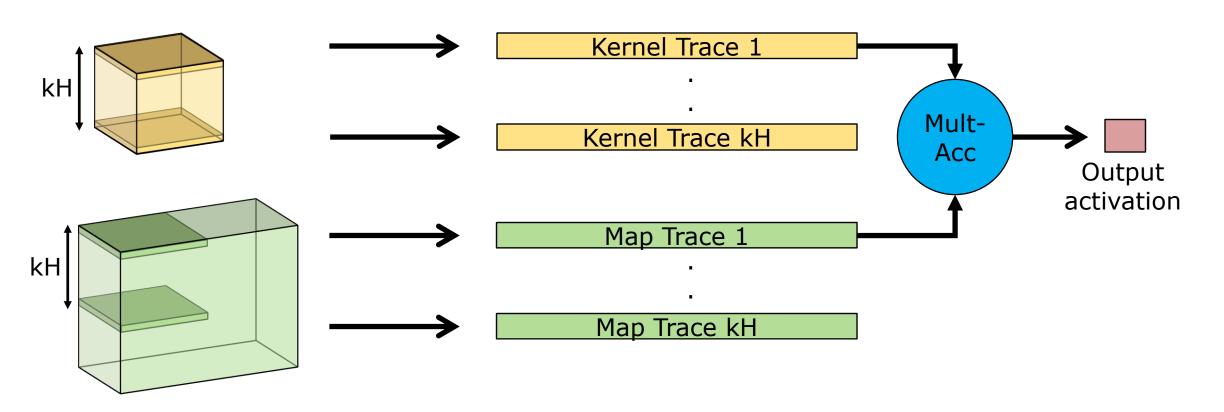


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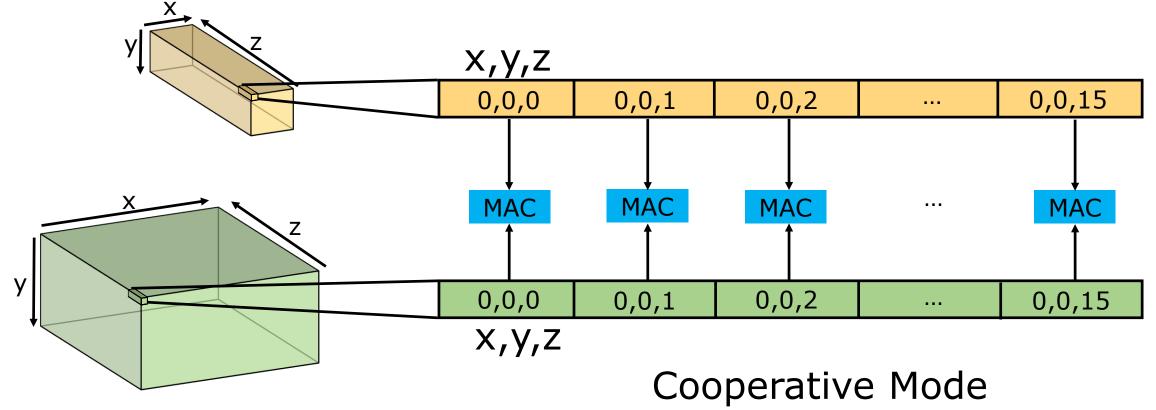
Data Organization



Data Organization

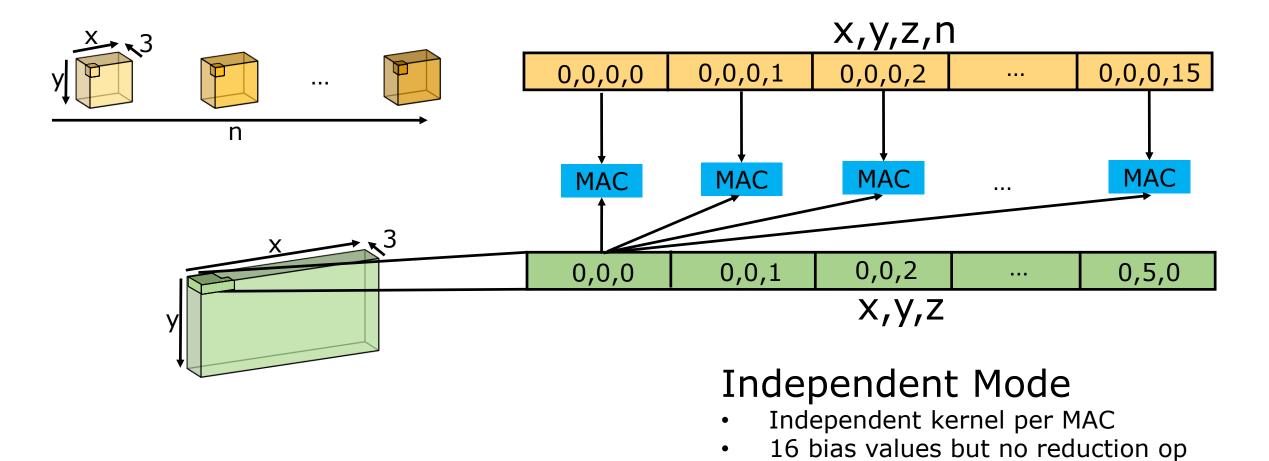


Intra-map, Intra-activation



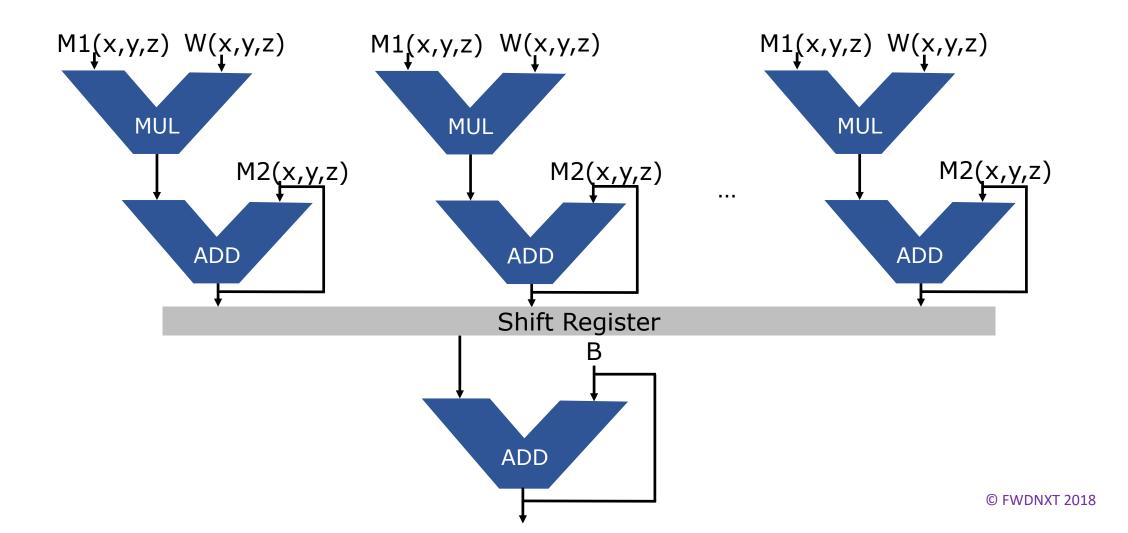
- Kernel shared by all MACs
- Single bias but need to reduce partials

Inter-map



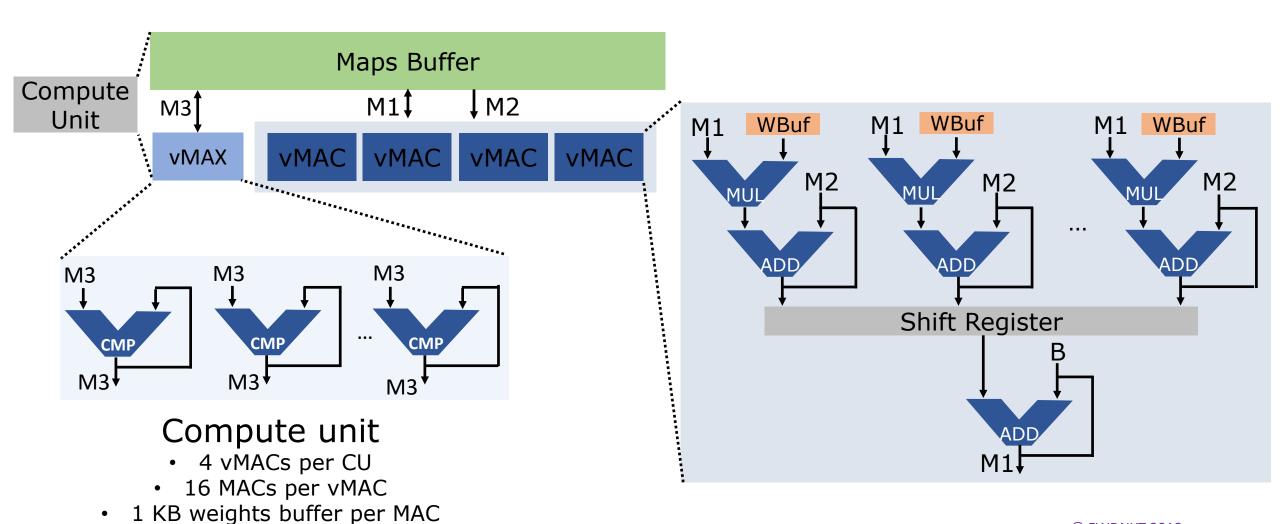
© FWDNXT 2018

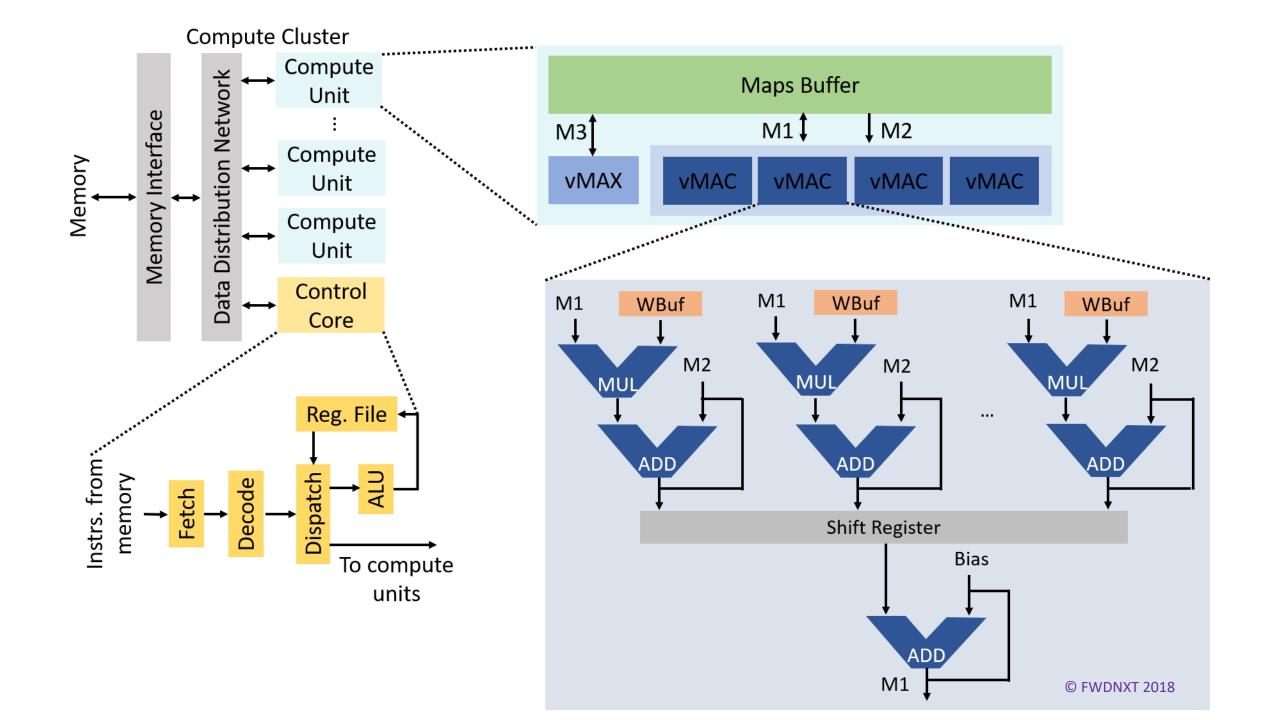
Vector Multiply-Accumulate (vMAC)



Scaling Up with Compute Units

64 KB maps (double) buffer per CU





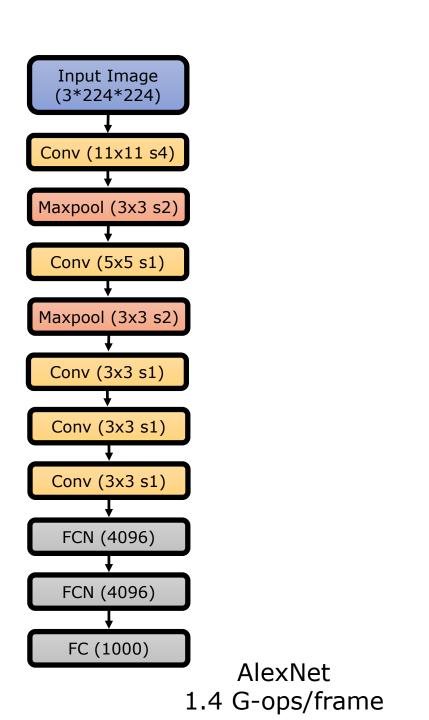
Types of Parallelism Revisited

Intra-map, intra-activation A vMAC in COOP mode vMACs across CUs Intra-map, inter-activation Inter-map vMACs within a CU Inter-layer vMACs across clusters © FWDNXT 2018

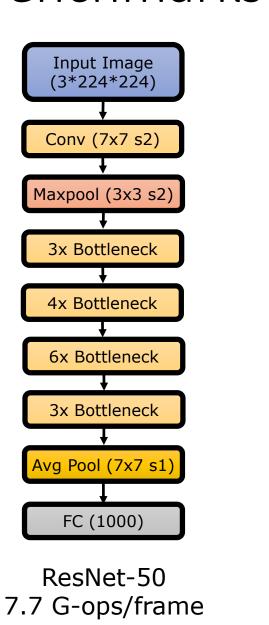
System Specifications

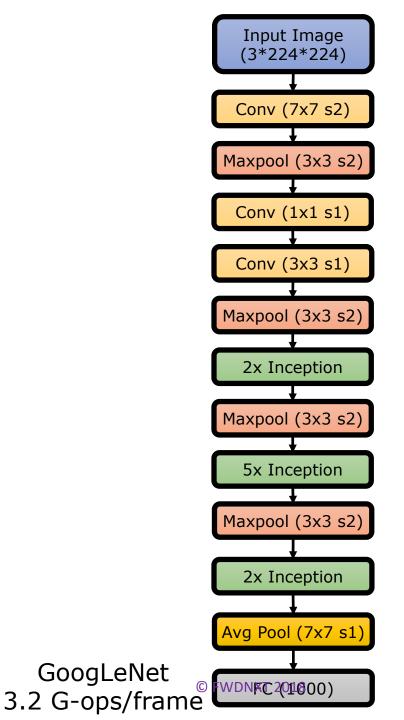
Host CPU	2x ARM Cortex-A9 @800 MHz
Accelerator cores	256 MAC units @ 250 MHz
Peak Throughput	128 G-ops/s
Memory	1GB DDR3 @ 533 MHz
Memory B/W	4.2 GB/s
Power (Board)	12 W
Power (Zynq + mem)	7 W



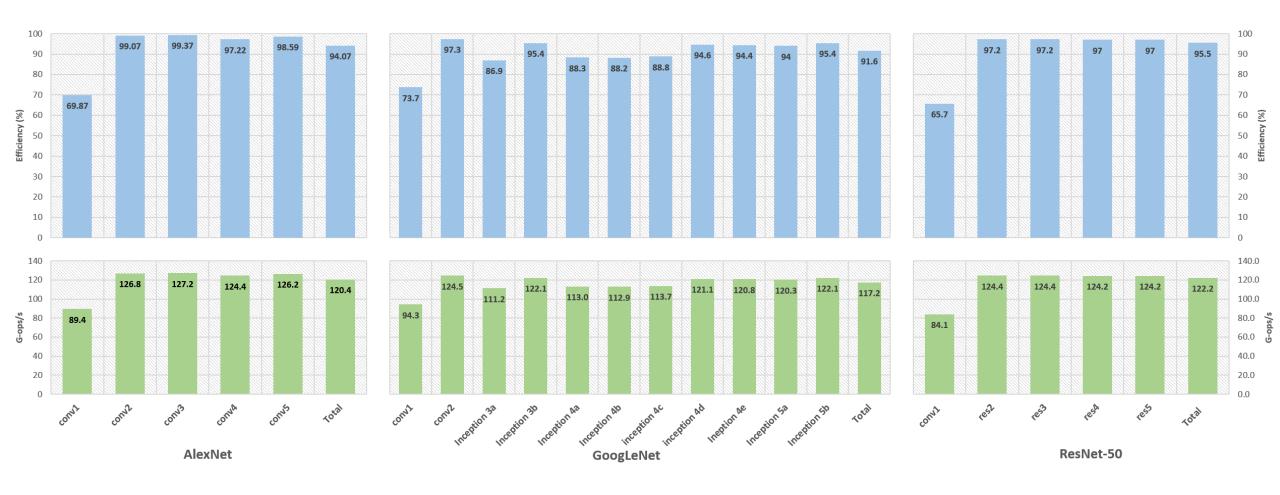


Benchmarks



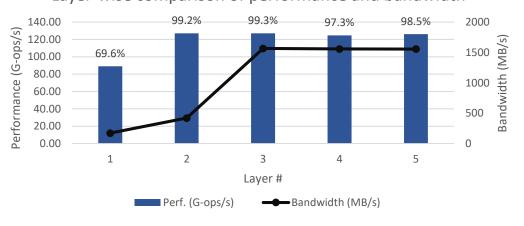


Performance

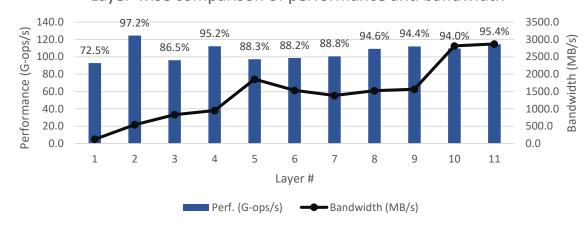


Comparison of Perf. and B/W

AlexNet Layer-wise comparison of performance and bandwidth



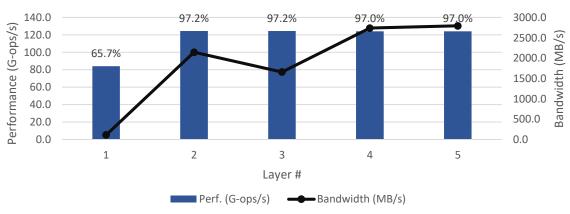
GoogLeNet Layer-wise comparison of performance and bandwidth



© FWDNXT 2018

ResNet-50

Layer-wise comparison of performance and bandwidth



Classification Results (top-5)



ambulance, minivan, minibus, golfcart, motor scooter



car, motorcycle, bicycle, watch, shoe



jaguar, dalmatian, banded gecko, leopard, bonnet



lionfish, jellyfish, sea slug, sea anemone, chambered nautilus



koala, wombat, sloth bear, mongoose, madagascar cat



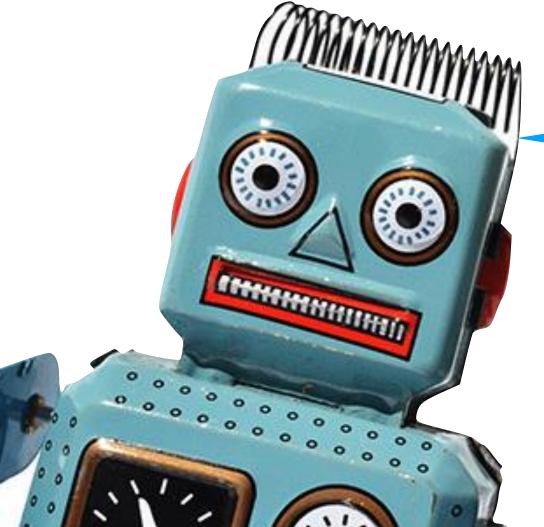
plastic bag, cauliflower, broccoli, swab, zucchini



motorcycle, bicycle, car, toy, watch



lion, cougar, hippopotamus, chimpanzee, book jacket



Thank you



