Riptide: Techniques for Fast End-to-End Binarized Networks

Josh Fromm

Original Data uint32

2 0 1 3

0 0 1 1

Bitplanes uint1

 0
 0
 1
 1

 1
 0
 0
 1

0 0 1 1

Bitpacked Data uint4

Bitserial Dot Product

 $1 \times popcount(3\&3) + 2 \times popcount(3\&9) = 4$

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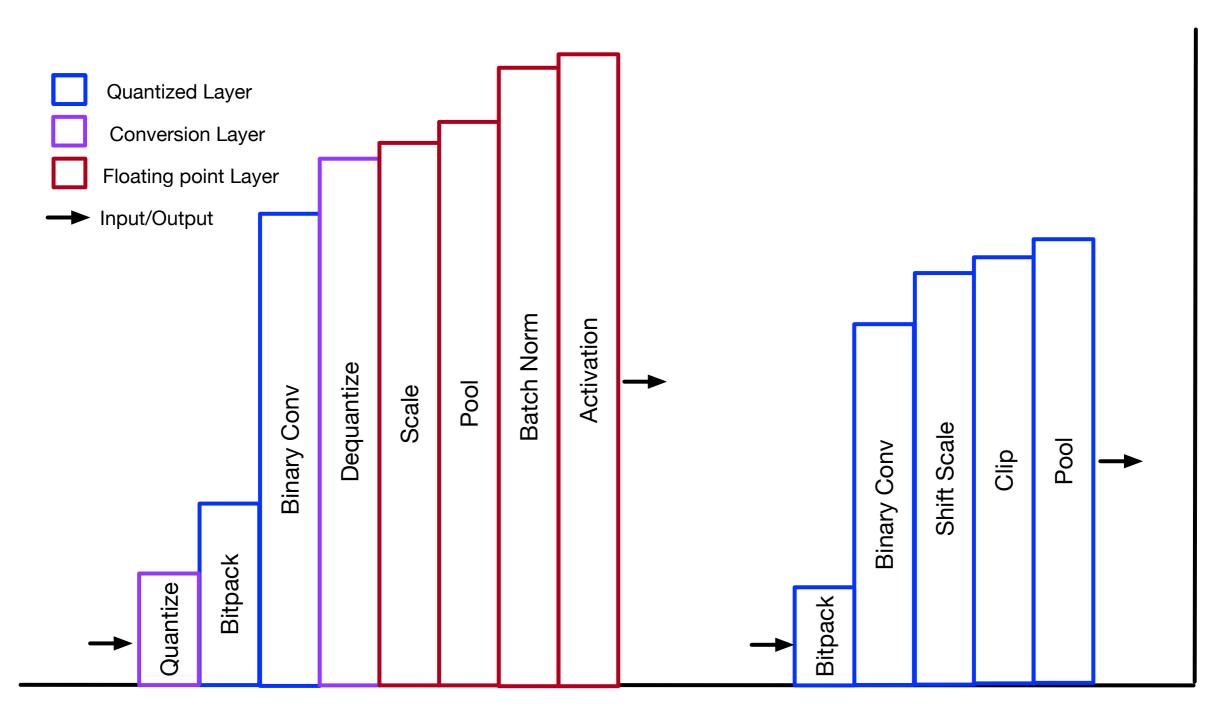
Bitpacked Data uint4

Bitserial Dot Product

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- Replaces 32-bit values with 1 or 2 bits
- Up to 32X compression
- Up to 48X (hypothetical) speedup

Fused Glue

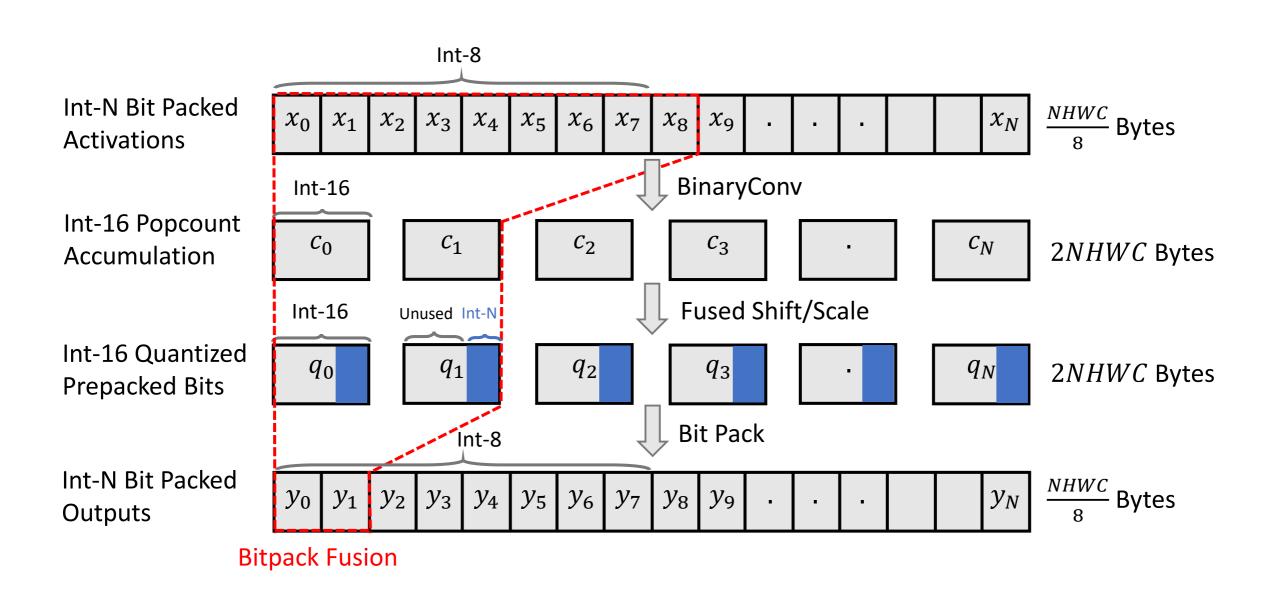


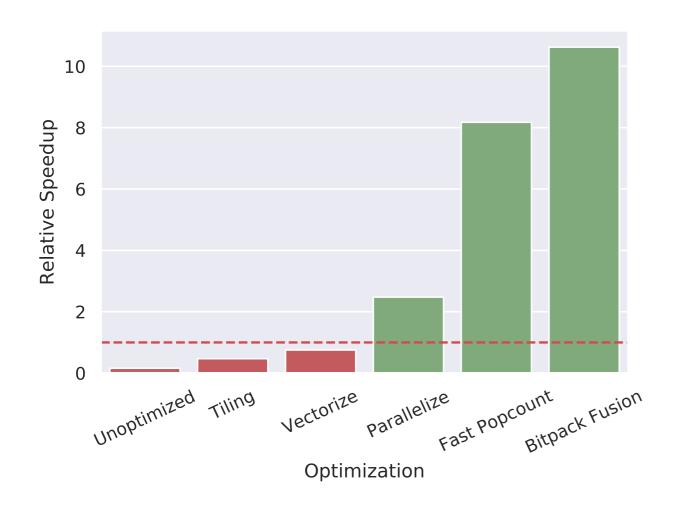
Cumulative Number of Operations

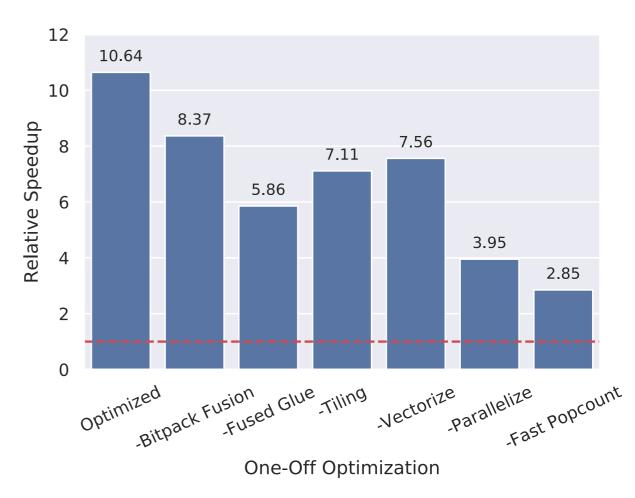
Traditional

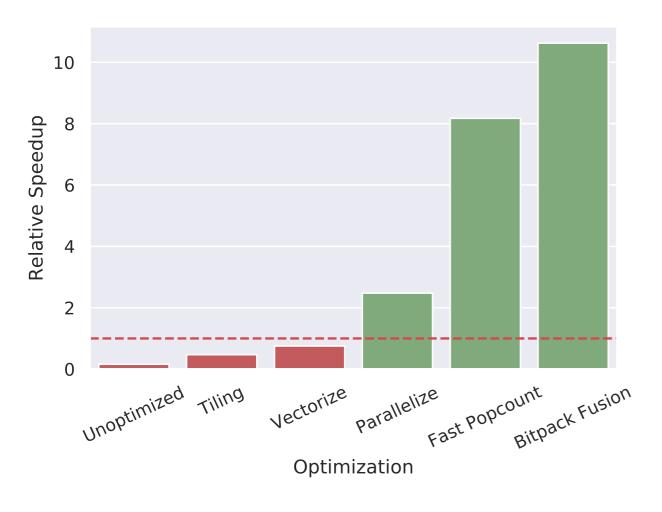
Fully Binarized (ours)

Bitpack Fusion



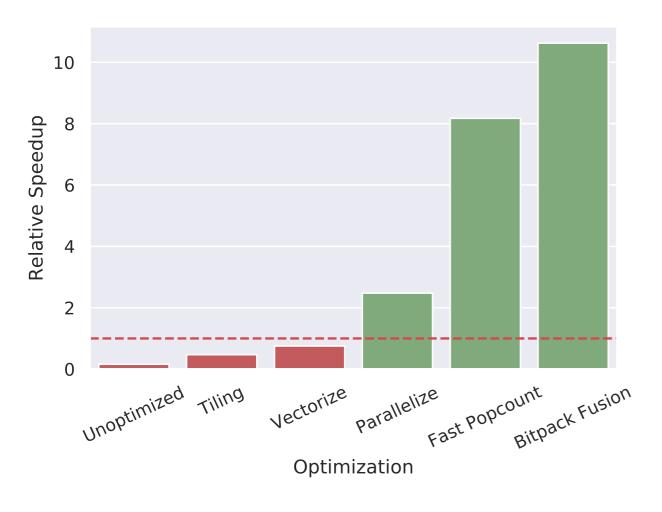






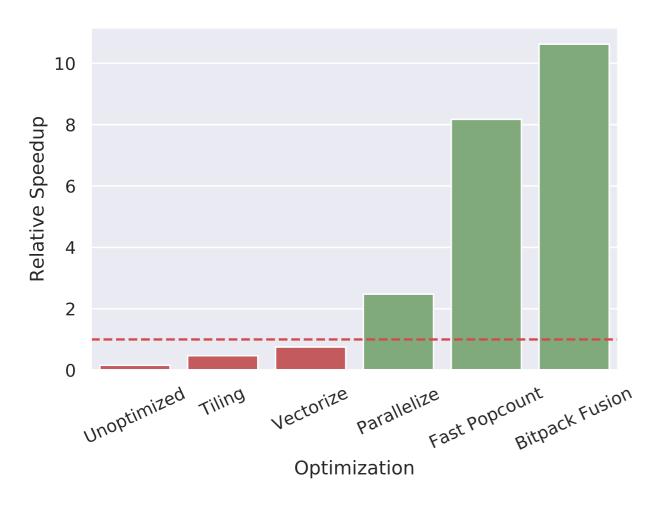


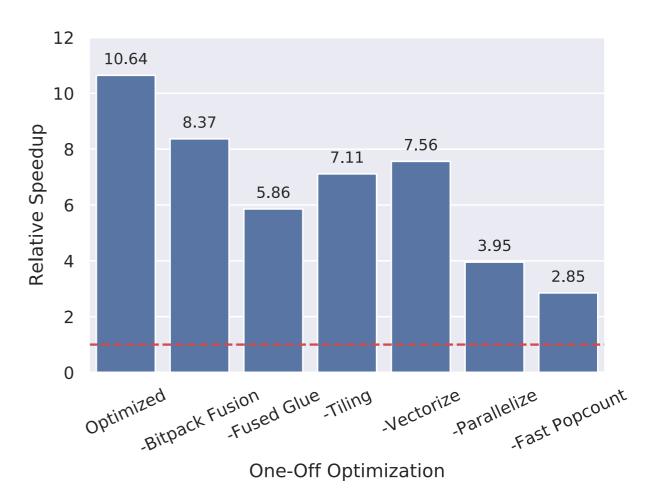
Up to 10X speedup



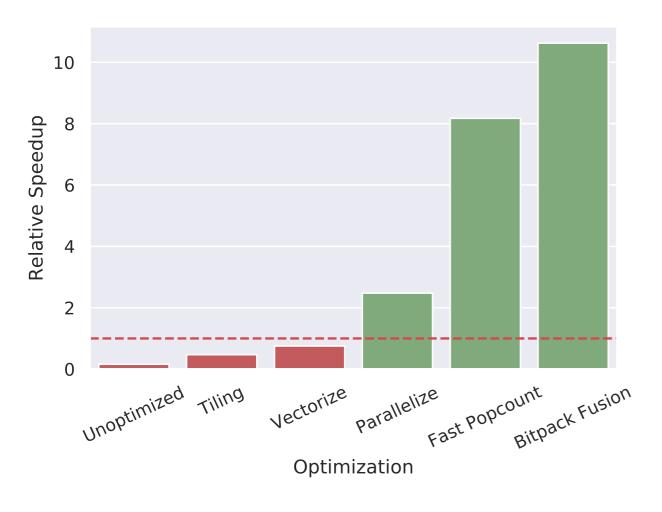


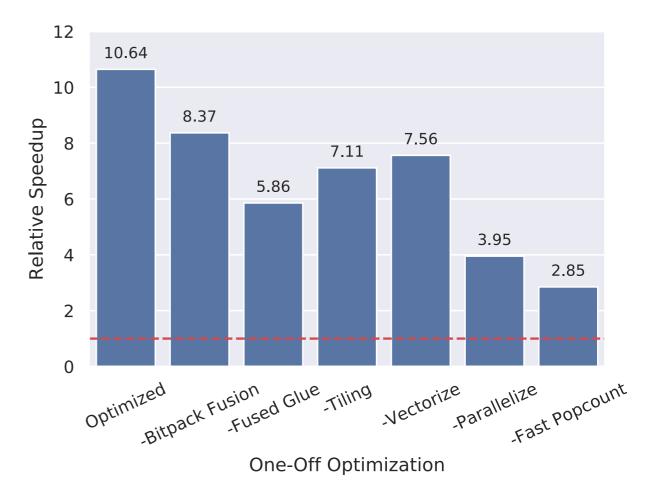
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- Memory savings from fused schedule yield 1.3X speedup





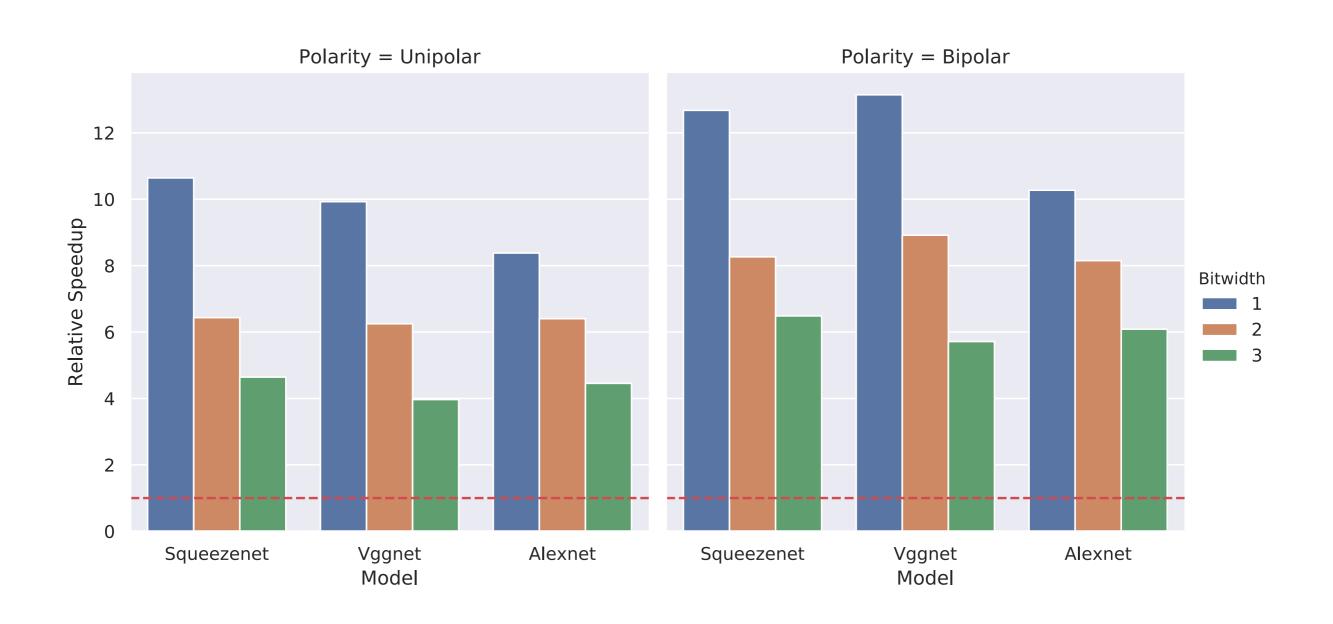
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- Up to 10X speedup
- Memory savings from fused schedule yield 1.3X speedup
- All optimizations have significant impact
- Fused glue offers a 2X speedup

All Speedups



Accuracy / Runtime

	Model	Name	1-bit	2-bit	3-bit	full precision			
	ImageNet top-1 accuracy / Runtime (ms)								
1	AlexNet	Xnor-Net [48]	$44.2\%\ /$	— / —	— / —	56.6% / —			
2	AlexNet	BNN [12]	27.9% / —	— / —	— / —	— / —			
3	AlexNet	DoReFaNet [63]	$43.6\%\ / -\!-$	$49.8\%\ / -\!-$	$48.4\%\ / -\!-$	55.9% / —			
4	AlexNet	QNN [27]	$43.3\%\ / -\!-$	51.0% / —	— / —	56.6% / —			
5	AlexNet	HWGQ [4]	— / —	52.7% / —	— / —	58.5% / —			
6	VGGNet	HWGQ [4]	— / —	64.1% / —	— / —	69.8% / —			
7	AlexNet	Riptide-unipolar (ours)	$44.5\%\ /\ 150.4$	52.5% / 196.8	$53.6\%\ /\ 282.8$	$56.5\%\ /\ 1260.0$			
8	AlexNet	Riptide-bipolar (ours)	$42.8\%\ /\ 122.7$	$50.4\%\ /\ 154.6$	$52.4\%\ /\ 207.0$	$56.5\%\ /\ 1260.0$			
9	VGGNet	Riptide-unipolar (ours)	$56.8\% \ / \ 243.8$	$64.2\%\ /\ 387.2$	67.1% / 610.0	$72.7\%\ /\ 2420.0$			
10	VGGNet	Riptide-bipolar (ours)	54.4% / 184.1	$61.5\% \ / \ 271.4$	$65.2\%\ /\ 423.5$	$72.7\%\ /\ 2420.0$			
11	ResNet18	Riptide-unipolar (ours)	$47.9\%\ /\ 76.2$	$58.4\% \ / \ 112.0$	$61.8\%\ /\ 152.3$	$70.9\%\ /\ 380.8$			

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Available Open Source Soon!