

Energy-Efficienct AI Acceleration through Approximation

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Outline



- Motivation
- Background
 - Approximation by quantization
- Hardware approximation
 - Stochastic computing (SC)
 - SC for neural network (NN) inference
 - SC for NN training
- Summary



Motivation



```
sitingliu - python - 64×20
(base) sitingliu@Sitings-MacBook-Air ~ % python
Python 3.9.13 (main, Aug 25 2022, 18:29:29)
[Clang 12.0.0] :: Anaconda, Inc. on darwin
Type "help", "copyright", "credits" or "license" for more inform
ation.
>>> 0.1+0.2
```



Motivation



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ation.
>>> 0.1+0.2
0.30000000000000004
>>>
```



Number Representation



- Fractions
 - Floating-point numbers (IEEE 754 standard single precision)

1 sign bit		t 8 bits	23 bits
	S Exponent		Significand/Mantissa

$$(-1)^S \times (1. Significand)_2$$

× $2^{Exponent-127}$

 Limited precision & rounding introduce accuracy loss

Motivation



- Digital systems are inherently inaccurate;
- Al applications can tolerate plenty inaccuracy;
 - Resilience from application and computation patterns.

App. no single correct answer

- E.g. recommendation systems, search query, etc.
- · Optimization problems such as neural network training.

Computation patters

- Map complex features (images, text) to several classes.
- Nonlinear functions have strong saturation effects.

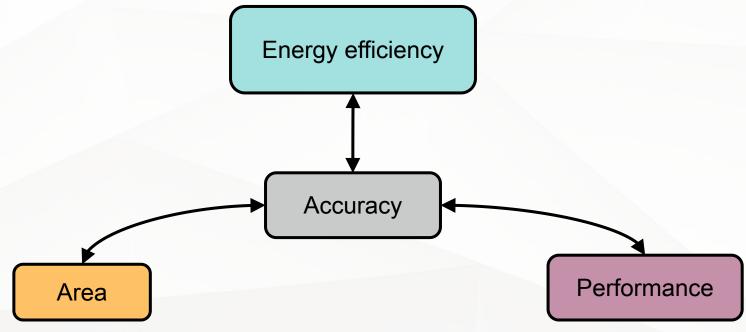


Motivation



- Digital systems are inherently inaccurate;
- Al applications can tolerate plenty inaccuracy;
- Trade accuracy for energy efficiency, performance, area, etc. through

Approximation





Approximation at Different Levels



Software/model

- Pruning
- Quantization
- Distillation
- Low-rank approximation
- •

Hardware/architecture

- Analog computing
- Approximate arithmetic circuits
- Stochastic computing
- •



Approximation at Different Levels



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Quantization—Motivation



 Rough energy numbers (in 45-nm technology node) from "Computing's Energy Problem, M. Horowitz, ISSCC, 2014"

INT				
ADD				
8 bit	0.03 pJ			
32 bit	0.1 pJ			
MULTI				
8 bit	0.2 pJ			
32 bit	3 pJ			

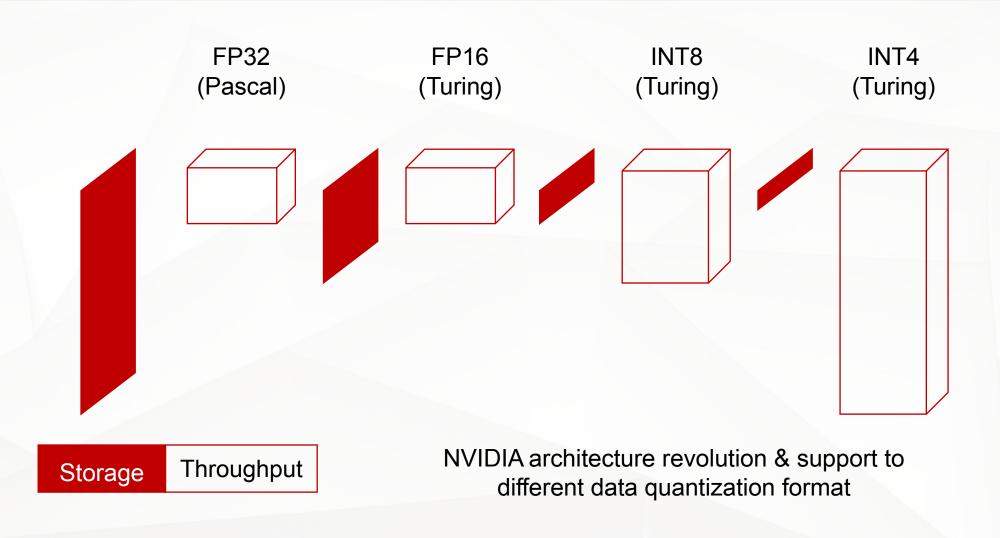
FP	
FADD	
16 bit	0.4 pJ
32 bit	0.9 pJ
FMULTI	
16 bit	1 pJ
32 bit	4 pJ

Memory		
Cache	(64 bit)	
8 KB	10 pJ	
32 KB	20 pJ	
1 MB	100 pJ	
DRAM	1.3-2.6 nJ	



Quantization—Hardware Implication

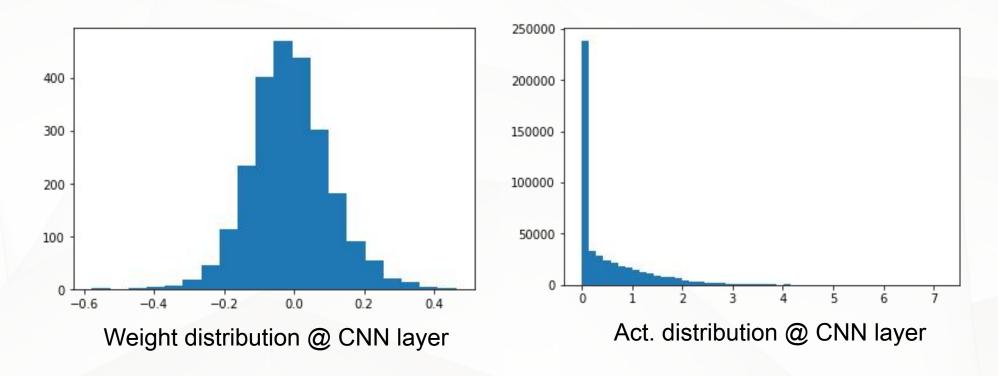






Quantization—Conventinoal Method





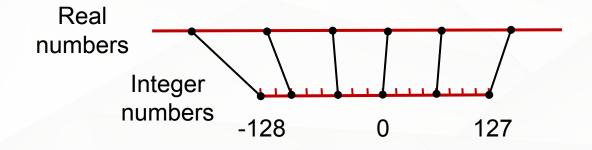
Scale, (shift) and round

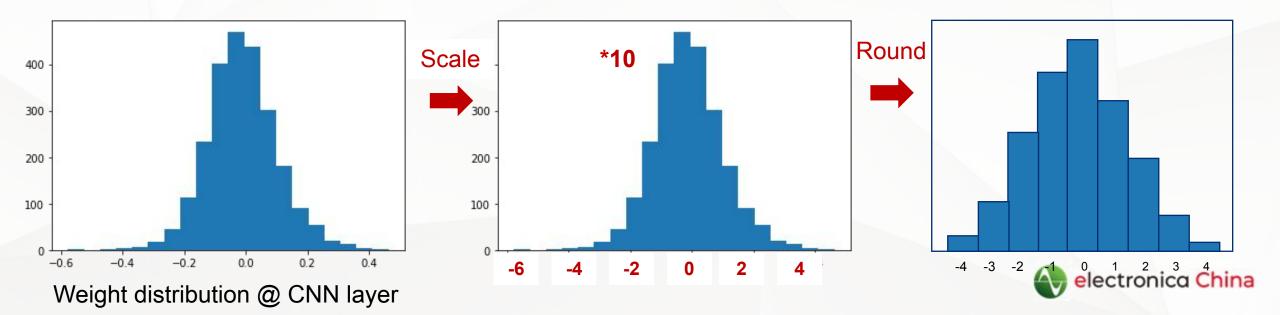


Quantization—Scale & Round



Floating-point → integer arithmetics

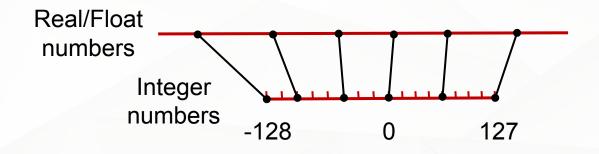




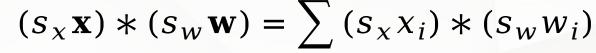
Quantization—Arithmetic Change



Floating-point → integer arithmetics



$$\mathbf{x} * \mathbf{w} = \sum x_i * w_i \qquad \longrightarrow \qquad (s_x \mathbf{x}) * (s_w)$$





 $\operatorname{round}(s_{\chi}\mathbf{x}) * \operatorname{round}(s_{w}\mathbf{w}) = \sum \operatorname{round}(s_{\chi}x_{i}) * \operatorname{round}(s_{w}w_{i})$



Quantization—Push the Limits



- Floating-point → INT4/FP8/FP4/Binary
 - But difficult to build a single piece of hardware to support all formats with different bit widths
- Another dimension: progressive precision in stochastic computing



Stochastic computing (SC) uses serial binary bits to represent a value.



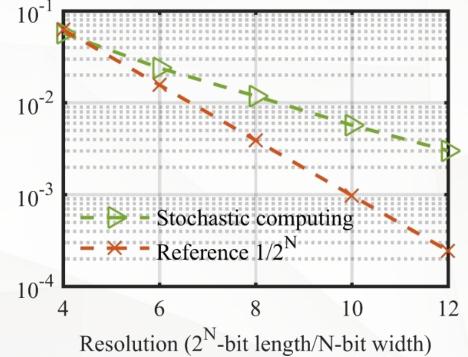
Stochastic Computing & Progressive Precision



Stochastic computing (SC) employs probability to encode a value.

4/8 = 0.5

 Since each bit is generated randomly, increasing the sequence length improves the accuracy.



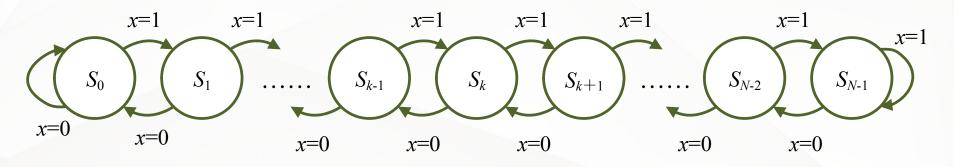


Stochastic Computing—Basic Elements



Simple logics gates perform complex arithmetics

Stochastic multiplier (AND gate)

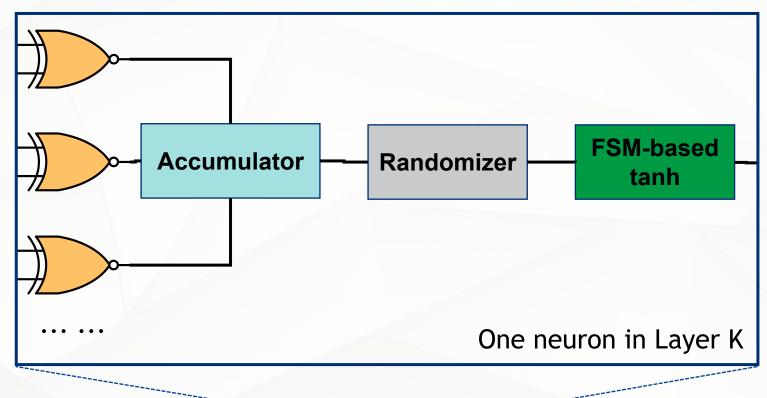


FSM implements tanh function in SC



An SC based-Neural Network Implementation





Model	Bit flip rate	Accuracy
MLP	0	97.77
	0	97.71
CC MI D	1	97.33
SC-MLP	5	96.92
	10	94.84

Bipolar		
stochastic multipliers	Layer 1	Layer 2

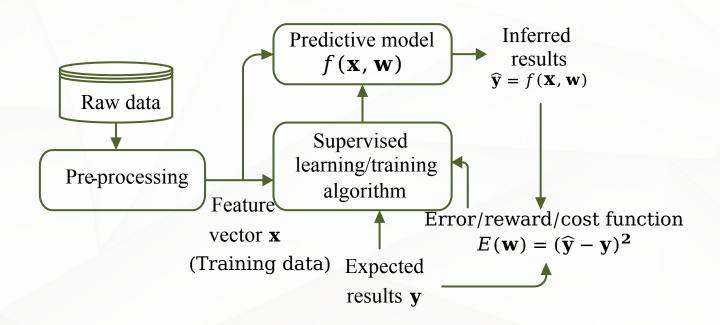
.. ... Layer N

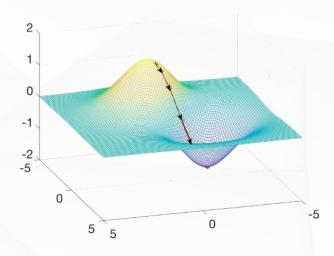


Stochastic Computing for NN Training



Gradient descent (with momentum) as optimizer





Gradient descent (GD) searching for local minimum.

• The optimization result is an accumulation of multiple steps of the gradients $g_i = \nabla E(\mathbf{w}_i)$.



SC-GDC Design

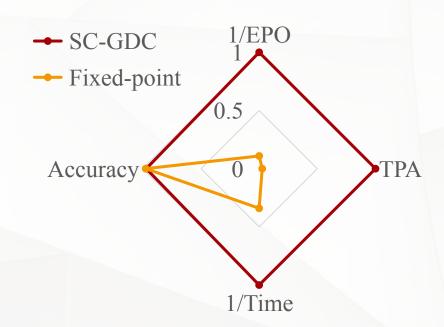


- Implement the iterative accumulation $w = \sum g_k$ in SC.
- g_k is stochastically quantized to -1/0/+1 and accumlated by a counter.



Hardware evaluation of the signed SC-GDC array training a 784-128-128-10 NN

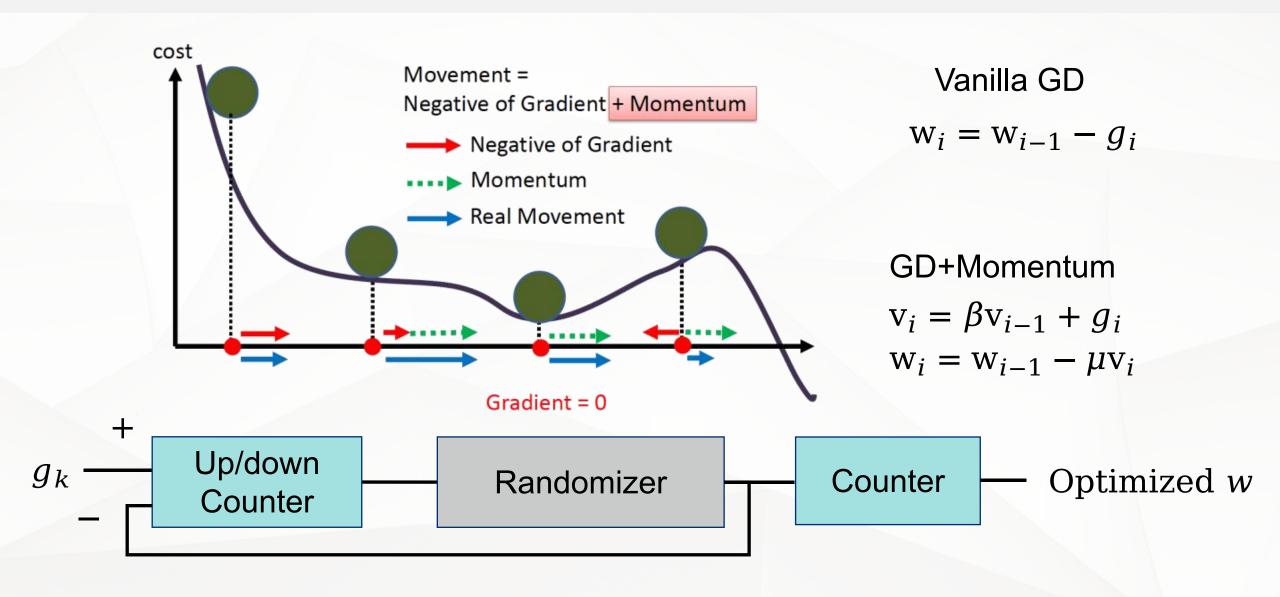
Metrics	SC-GDCs	Fixed-point
Step size	2^{-10}	2^{-10}
Epochs	20	20
Min. time (ns)	$1.6 imes 10^6$	$4.7 imes10^6$
EPO (fJ)	1.2×10^{7}	1.1×10^{8}
TPA (images/ $s/\mu m^2$)	5.7×10^{1}	1.5
Aver. test Accu.	97.04%	97.49%





SC-GDM Design

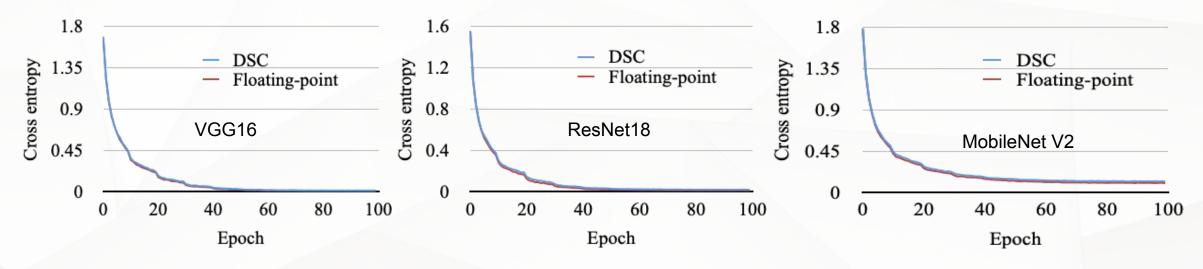




Experiments & Results



 Can train more complex NN architecture such as VGG16, ResNet18 and MobileNet V2 with CIFAR10 dataset.



Test accuracy (%)	VGG16	ResNet18	MobileNetV2
SC-GDM	90.23	91.36	88.51
Floating-point	90.55	91.85	88.82

Take-aways



- With numerical approximation, the computation efficiency can be improved;
- Stochastic computing (SC) is a numerical approximation technique that represents a value by the probability of a binary bit stream;
- SC is able to achieve bit-flip-resilient NN inference and energy-efficient and high-performance training;
- It achieves a higher energy efficiency and performance compared with traditional computing paradigm.

Thanks for your attention!



Q & A