

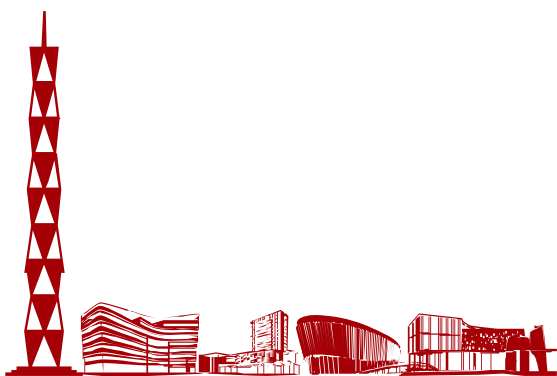
Can Stochastic Computing Truly Tolerate Bit Flips?

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Outline

- Background: stochastic computing & stochastic circuits
- Bit-flip model and bit-flip error recovery of stochastic circuits
- Experiments & results
- Application
- Conclusion



Background: stochastic computing

- In stochastic computing (SC), information is encoded and processed by streams of random bits. The probability of 1 (p) in the bit stream is used to represent the value (x).

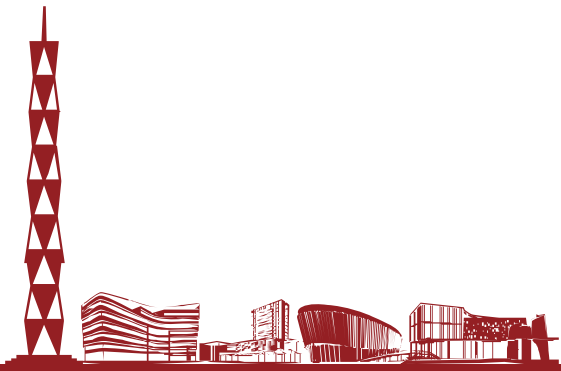
- Binary representation: $(01000101)_2 \rightarrow 69$

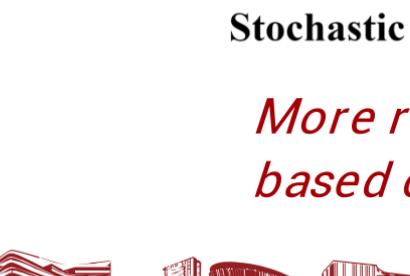
- Stochastic computing (SC): $01000101 \left\{ \begin{array}{l} 3/8 \text{ (unipolar)} \\ -1/4 \text{ (bipolar)} \end{array} \right. \quad x = 2p - 1$

$(00000101)_2 \rightarrow 5$

$00000101 \left\{ \begin{array}{l} 2/8 \text{ (unipolar)} \\ -2/4 \text{ (bipolar)} \end{array} \right.$

More resilient to bit flip errors





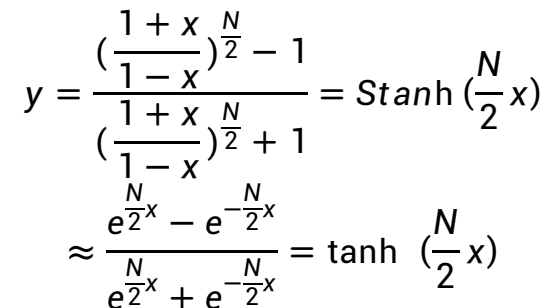
Stochastic FSI

*More resilient
based on*



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Bit flip models of stochastic sequences

- Assume bit flips are Bernoulli events with probability P_E and they are independent
- Stochastic sequence with probability P_X , representing x
- Probability considering bit flips P'_X

$$x' = P'_X = P_X(1 - P_E) + (1 - P_X)P_E = x + (1 - 2x)P_E. \text{ (unipolar)}$$

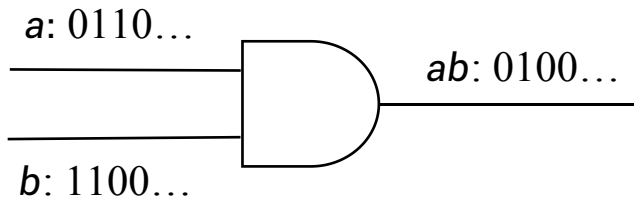
$$x' = (1 - 2P_E)x. \text{ (bipolar)}$$



Each bit has a
probability of
 P_E to flip

Bit flip models of stochastic multipliers

■ Unipolar stochastic multipliers

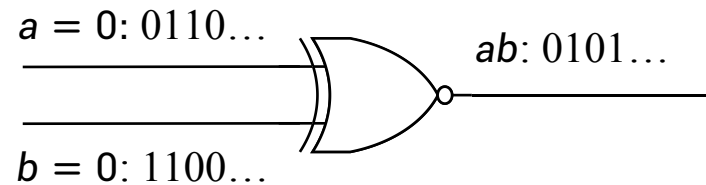


$$\begin{aligned} a'b' \\ &= (1 - 2P_E)^2 ab \\ &+ (1 - 2P_E)P_E(a + b) + P_E^2. \end{aligned}$$

■ Error recovery

Given $a'b'$, **cannot** directly compute ab

■ Bipolar stochastic multipliers



$$a'b' = (1 - 2P_E)^2 ab$$

■ Error recovery

Given $a'b'$, directly compute ab by a linear transformation

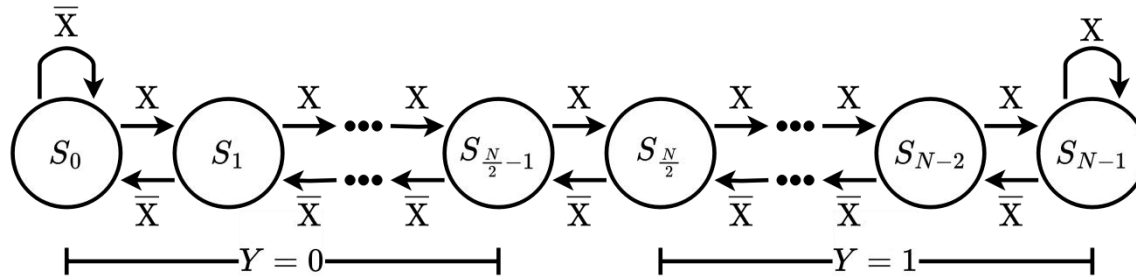
$$\text{Recovery value} = \frac{\text{Distorted value}}{(1 - 2P_E)^2}$$

Not possible in conventional arithmetic!



Bit flip models of stochastic FSM

- FSM-based stochastic circuits, Stanh as an example



Stochastic FSM implementing tanh function

$$y' = \text{Stanh}\left[\frac{N}{2}(1 - 2P_E)x\right]$$

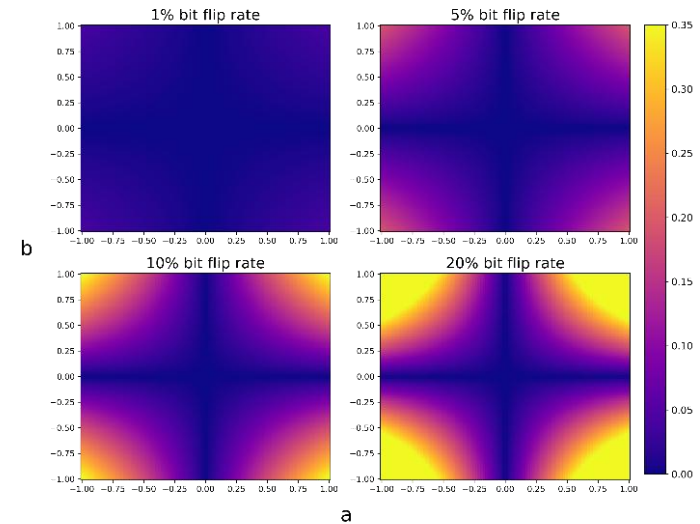
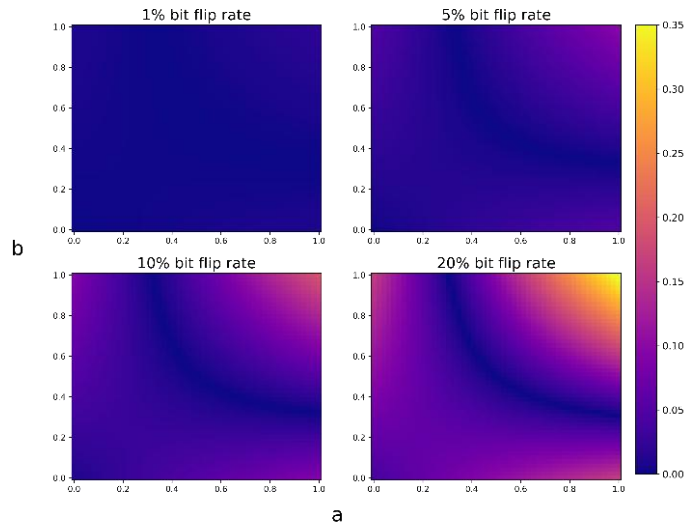
- Error recovery

$$y = \frac{\frac{1 + y'^{\frac{1}{1-2P_E}}}{1 - y'} - 1}{\frac{1 + y'^{\frac{1}{1-2P_E}}}{1 - y'} + 1}$$

Not possible in
conventional
arithmetic!

Experiments & Results

■ Verification of the bit-flip models for stochastic multipliers



Absolute error of stochastic multiplication with different bit flip rates in the (left) unipolar representation and (right) bipolar representation.

Error is 0 when

$$a = b = 0.5 \text{ or } \frac{a}{2a-1} + \frac{b}{2b-1} = P_E (a \neq 0.5, b \neq 0.5)$$

Error is 0 when

$$a = 0 \text{ or } b = 0$$

Error recovery of bipolar stochastic multipliers

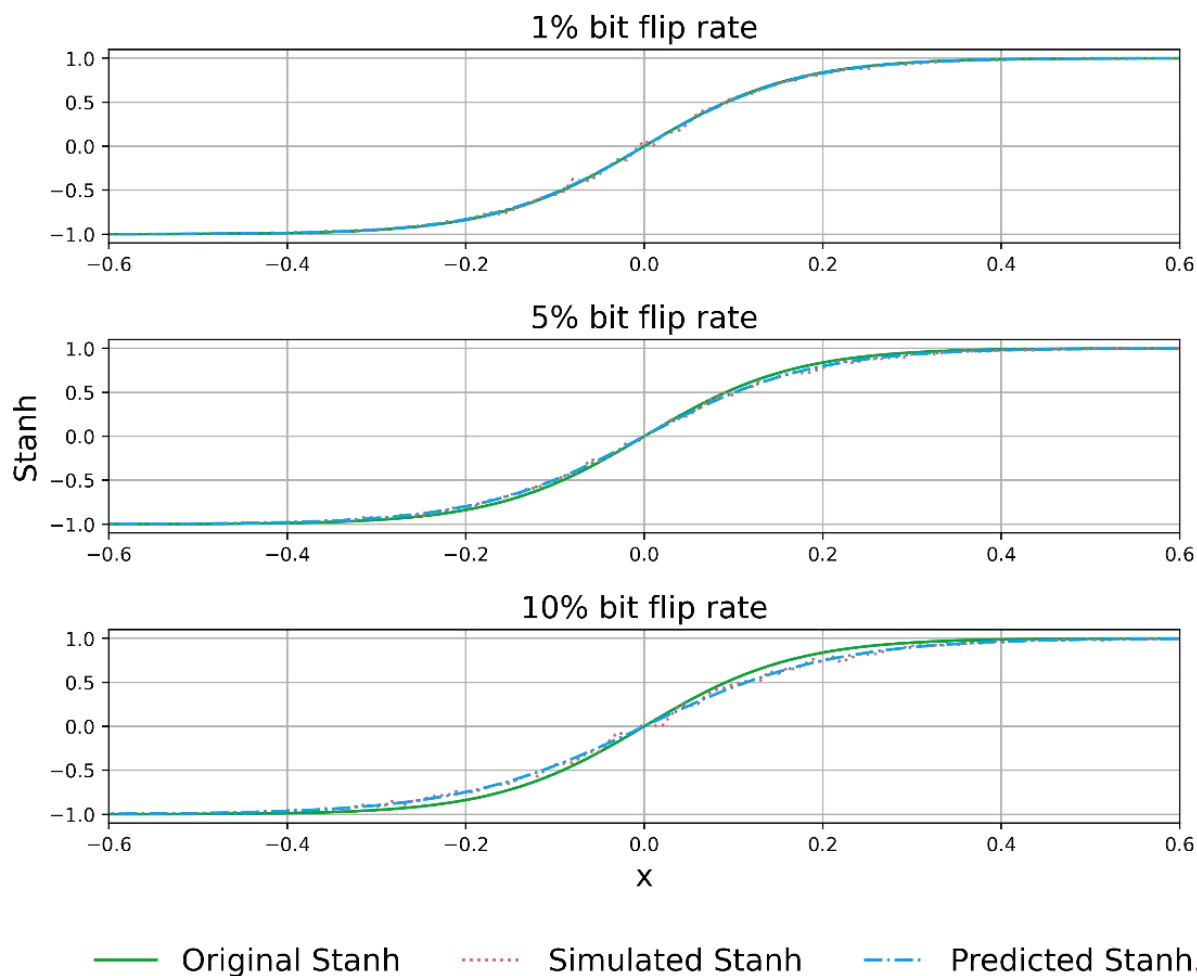
- Error recovery with different bit flip rates.
- 100 trials are performed, and RMSEs are measured.

Table 1. the RMSEs of stochastic multipliers before and after error recovery

Bit-flip rate	0%	1%	5%	10%
Distorted	0.0011	0.0164	0.0514	0.0941
Recovered	N/A	0.0011	0.0015	0.0021

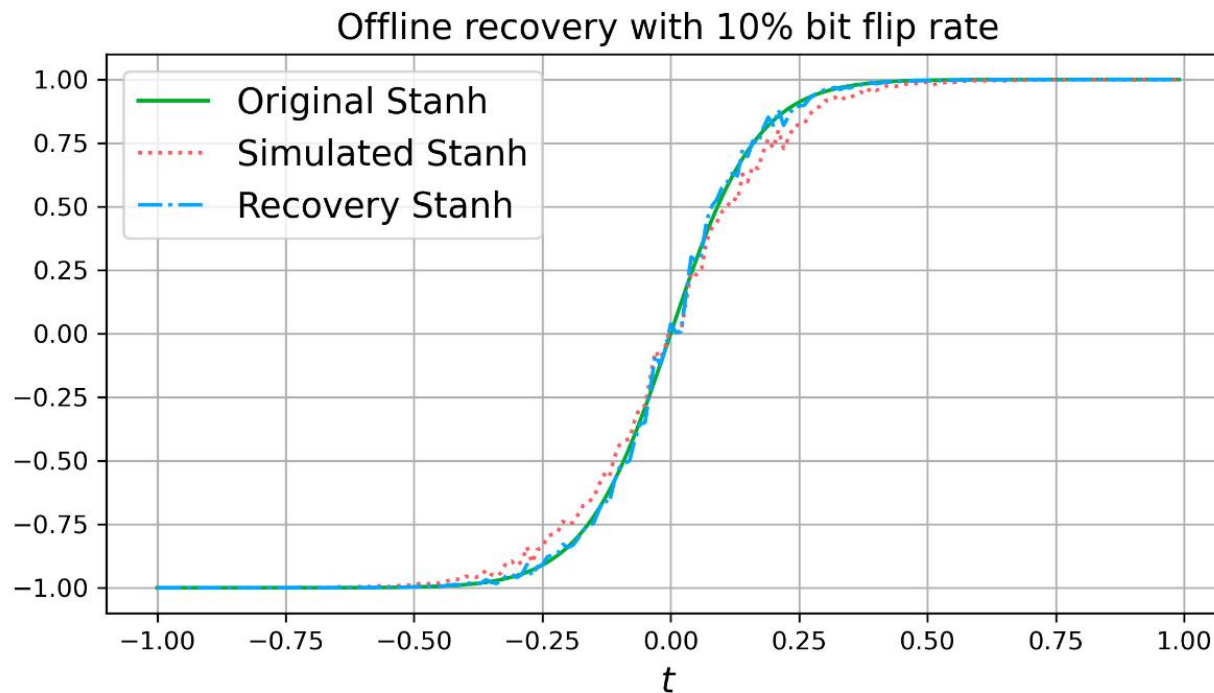
Stochastic FSM models

■ Verification of the bit-flip model



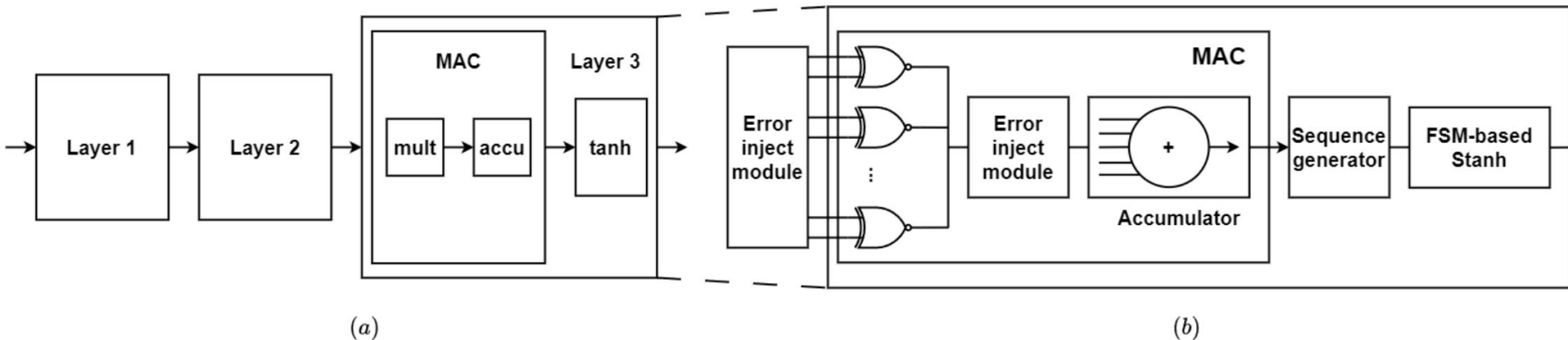
Stochastic FSM models (Cont'd)

■ Error recovery



Application in a multi-layer perceptron (MLP)

- A three-layer (784-256-256-10) MLP is implemented by the aforementioned SC components
- For each layer:
$$f_i(x_i) = \tanh(2(x_i \times w_i)),$$
- Equation for the entire network:
$$\text{output} = f_3(f_2(f_1(x))).$$



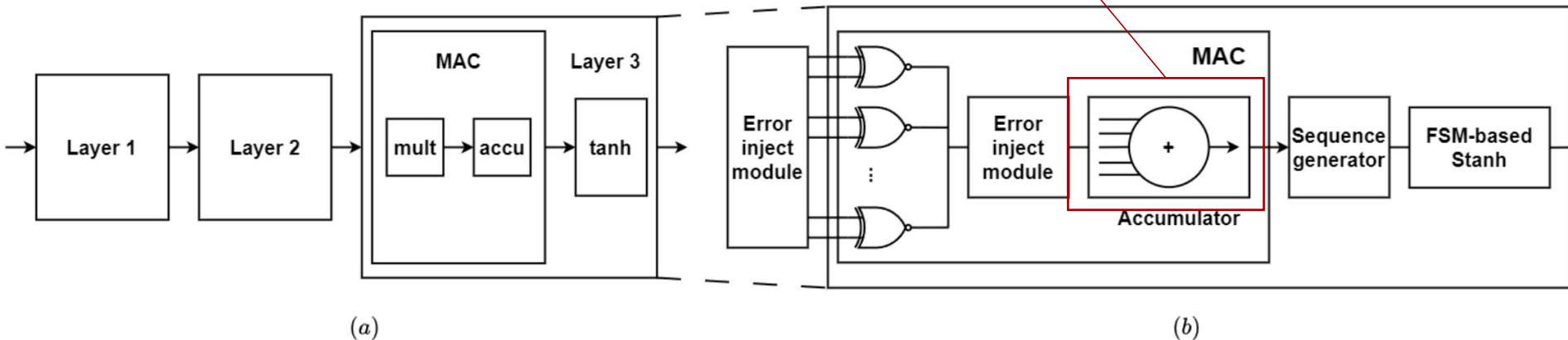
Error-free SC-based MLP circuit

SC-based MLP circuit with error injection

Error model of the SC accumulator

$$\mathbb{E}[\text{sum}] = \sum (2P_{x_k} - 1) = \sum x_k$$

$$\begin{aligned} \text{sum}^d &= \sum x'_k = \sum (1 - 2P_E) x_k \\ &= (1 - 2P_E) \text{sum} \end{aligned}$$



Error-free SC-based MLP circuit

SC-based MLP circuit with error injection

Error model of the entire MLP circuit

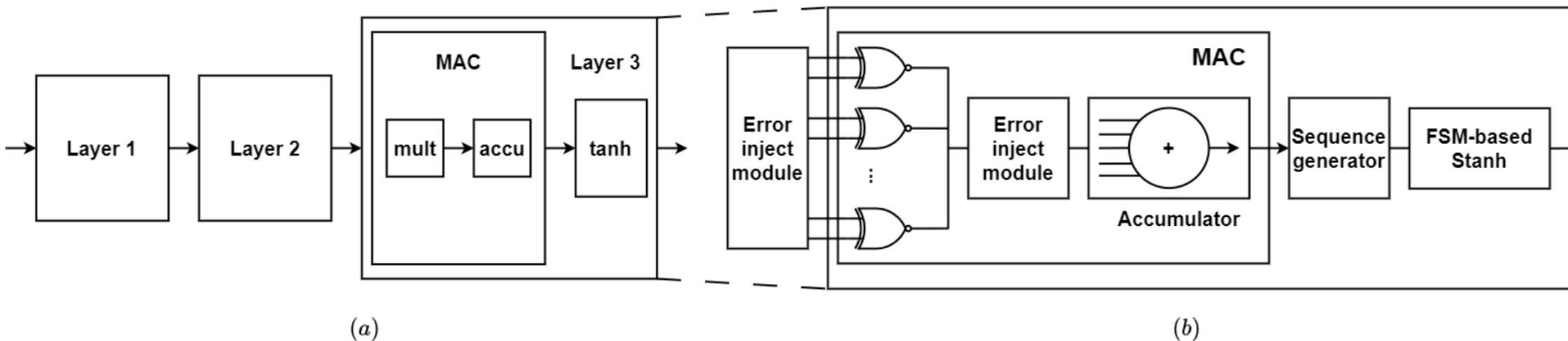
$$f_i(x_i) = \tanh(2(x_i \times w_i))$$

$$\text{output} = f_3(f_2(f_1(x)))$$

\Rightarrow

$$f_i^d(x_i) = \tanh(2(1 - 2P_E)^4(x_i \times w_i))$$

$$\text{output}^d = f_3^d(f_2^d(f_1^d(x)))$$



Error-free SC-based MLP circuit

SC-based MLP circuit with error injection

Verification of the bit flip model

- A three-layer (784-256-256-10) MLP is implemented by the aforementioned SC components on MNIST dataset.

Table 2: Accuracies of simulated and predicted results under different bit-flip rates

Model	Bit flip rate (%)	Simulated accuracy (%)	Predicted accuracy (%)
Pre-trained MLP	0	97.77	–
SC-MLP	0	97.71	97.77
	1	97.33	97.75
	5	96.92	97.48
	10	94.84	96.65



Error recovery of the SC-based MLP circuit

- An online layer-by-layer error recovery strategy is proposed
- For each layer:

$$f_i(x) = \frac{\frac{1 + f_i^d(x) \frac{1}{(1-2P_E)^4}}{1 - f_i^d(x)} - 1}{\frac{1 + f_i^d(x) \frac{1}{(1-2P_E)^4}}{1 - f_i^d(x)} + 1}$$

$$f_i(x_i) = \tanh(2(x_i \times w_i))$$

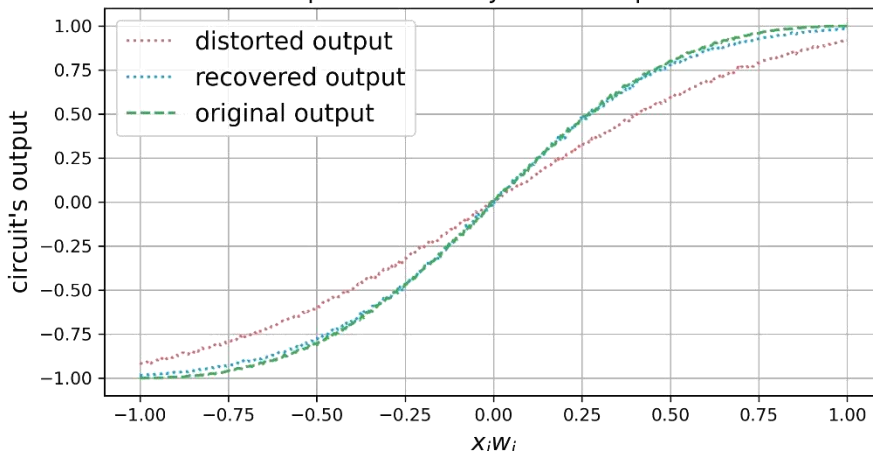
$$\text{output} = f_3(f_2(f_1(x)))$$

⇐

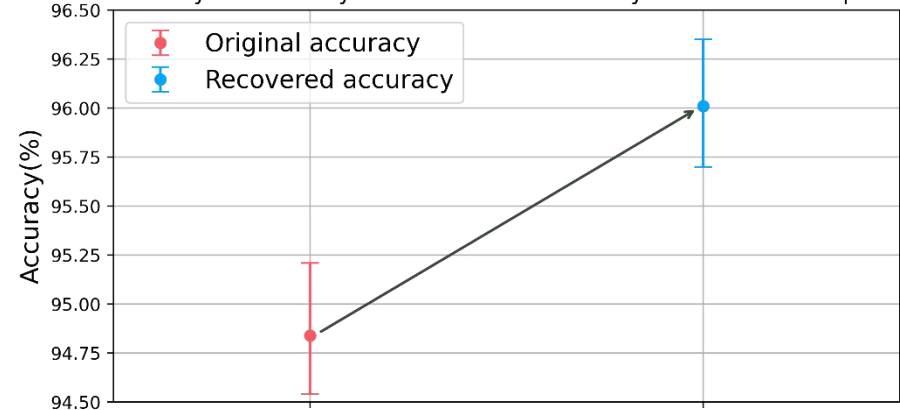
$$f_i^d(x_i) = \tanh(2(1 - 2P_E)^4(x_i \times w_i))$$

$$\text{output}^d = f_3^d(f_2^d(f_1^d(x)))$$

Different output of the SC system for input from -1 to 1.

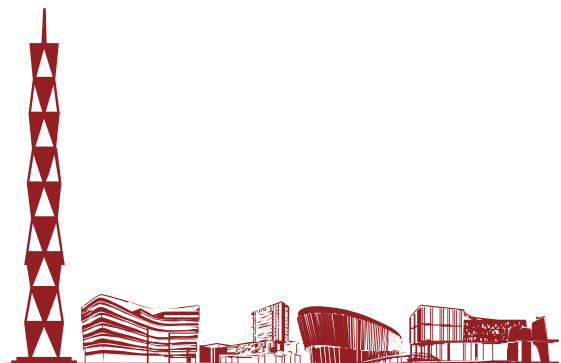


Accuracy of the SC system and online recovery under 10% bit-flip rate



Conclusion and future work

- Bit-flip models of stochastic sequences and multiple stochastic circuits are established;
- Formulation are conducted and bit-flip error recovery is proved to be possible on bipolar stochastic arithmetics;
- Bit-flip error recovery on independent SC components shows a high accuracy;
- An online layer-by-layer error recovery strategy is proposed to deal with an SC-based MLP circuit. The accuracy is partially recovered.
- More factors will be considered in the future work to build more accurate SC bit-flip models.



Thank you for your attention!

Q&A