

Tutorial: *p*-bits for Probabilistic Spin Logic

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Joint University Microelectronics Program



p-bits for Probabilistic Spin Logic (PSL)

Zhihong Chen



Supriyo Datta



- (1) Camsari et al., *Phys. Rev. X* (2017)
(Computing with p-bits)
- (2) Camsari et al., *IEEE EDL* (2017)
(Stochastic MTJ-based p-bit)
- (3) Camsari et al., *arXiv:1810.XXXX* (2018)
(Quantum Annealing with p-bits)
- (4) Camsari et al., *arXiv:1809.04028*(2018)
(Review of PSL)

Punyashloka
Debashis



Ramtin Zand
(UCF)



Ron DeMara
(UCF)



Sayeef Salahuddin
(UC Berkeley)



Anirudh
Ghantasala



Shuvro
Chowdhury



Behtash
Behin-Aein (GF)



Brian M Sutton



Zeeshan Pervaiz



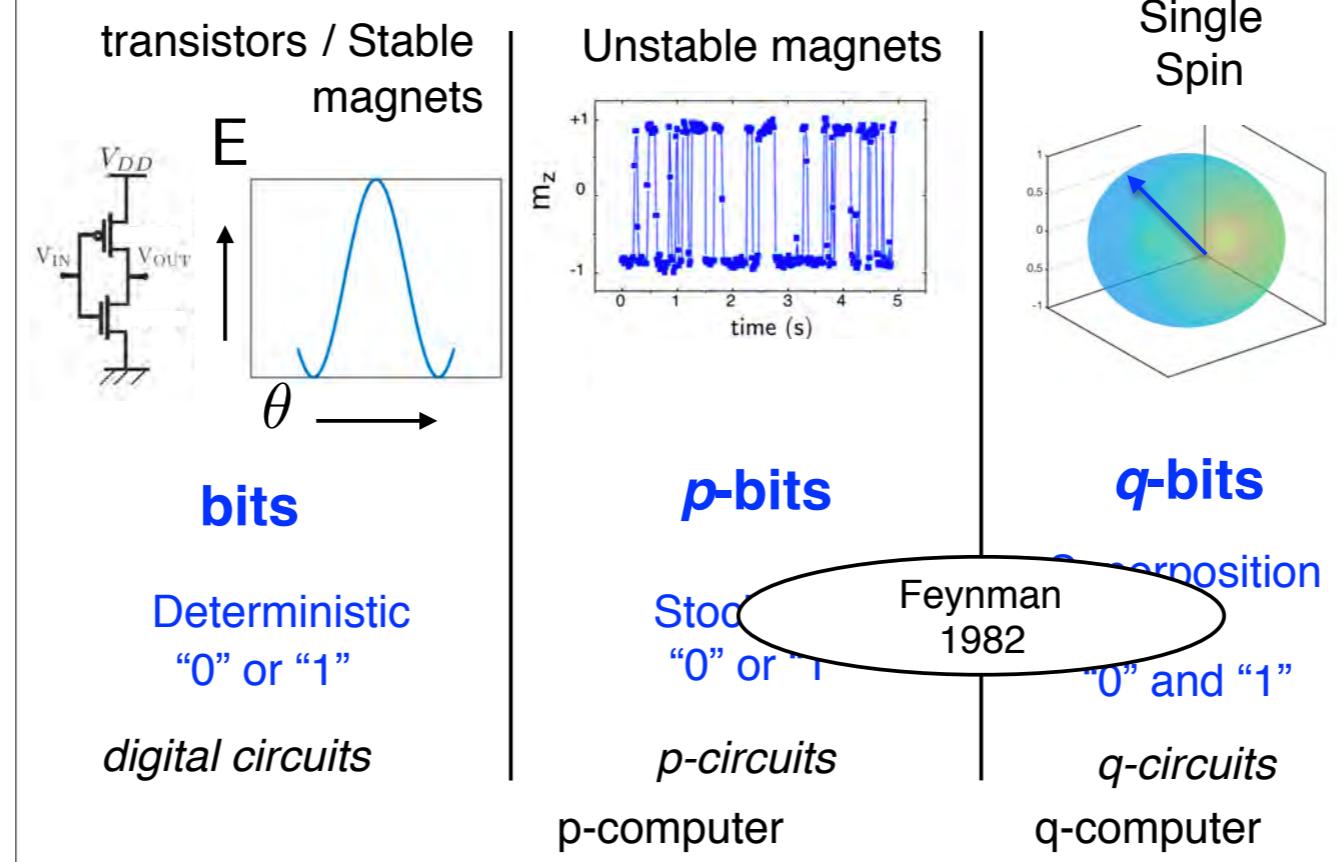
Rafatul Faria



Orchi Hassan



bits, p-bits and q-bits



Digital electronics .. notion of a "bit" , charge on a capacitor

Experimental data by Debasish / Chen

I will make a concrete p-bit to q-bit connection

Application space for p-bits

Quantum Computing inspired

Adder \leftrightarrow Subtractor

(1) Subset Sum Problem

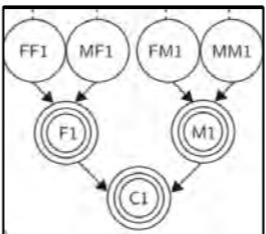
Multiplier \leftrightarrow Divider

(2) Prime Factorization

Camsari et al., *Phys. Rev. X*, (2017)

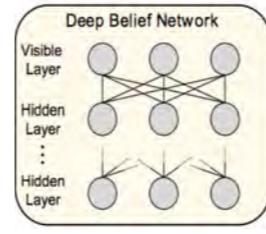
Machine Learning inspired

Bayesian Networks



Faria et al., *AIP Advances* (2018)

DBN's using p-bits



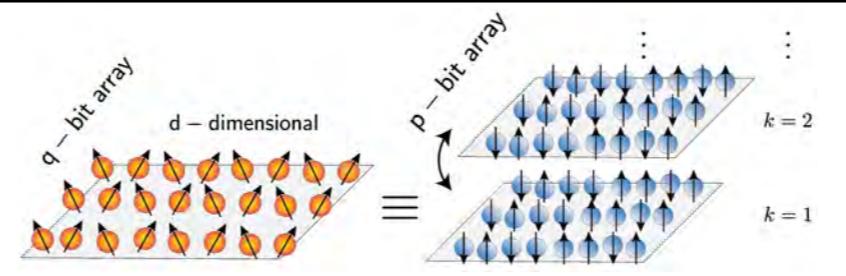
Zand et al.,
GLSVLSI (2018)

Combinatorial Optimization (e.g. TSP)



Sutton et al., *Sci. Rep.*, (2017)

Quantum circuits emulated with p-bits



Camsari et al., *arXiv:1810.XXXX*

- **How to build a p-bit?**
- Quantum Computing-inspired (Reciprocal Networks)
- Machine Learning-inspired (Reciprocal and Non-reciprocal Networks)

p-bit is a binary stochastic neuron

Hardware-specific features

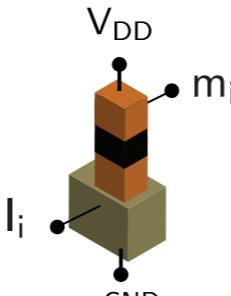
* **Hardware p-bit is a three-terminal device**

so that it can be interconnected as a transistor-like building block

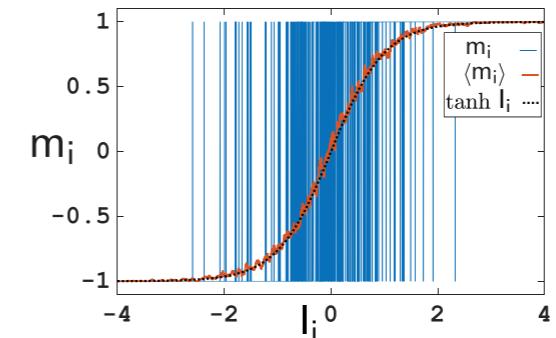
** **update order of p-bits:**
needs to be sequential or *asynchronous*

*** **speed of synapse:**
-needs to be much faster than *p-bit*

Need for hardware *p-bits*



p-bit



Basic Equations

$$m_i = \text{sgn}[\text{rand}(-1, +1) + \tanh(I_i)] \quad (1)$$

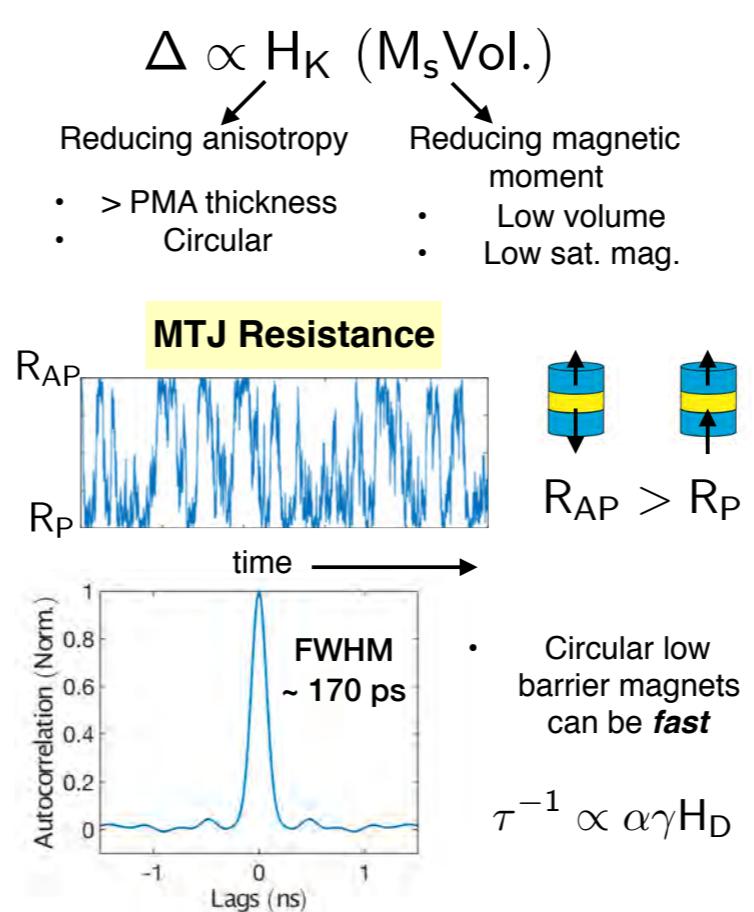
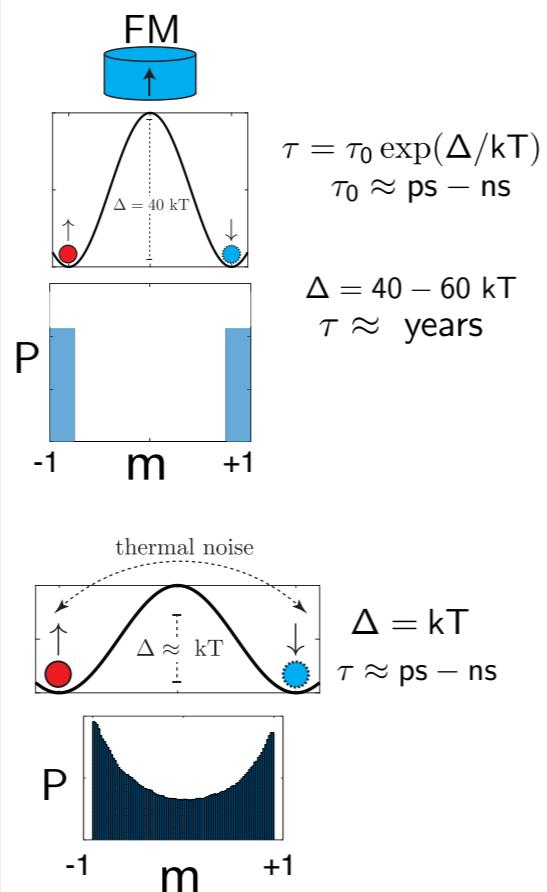
p-bit: Stochastic binary neuron

$$I_i = I_0 \left(\sum J_{ij} m_j + h_i \right) \quad (2)$$

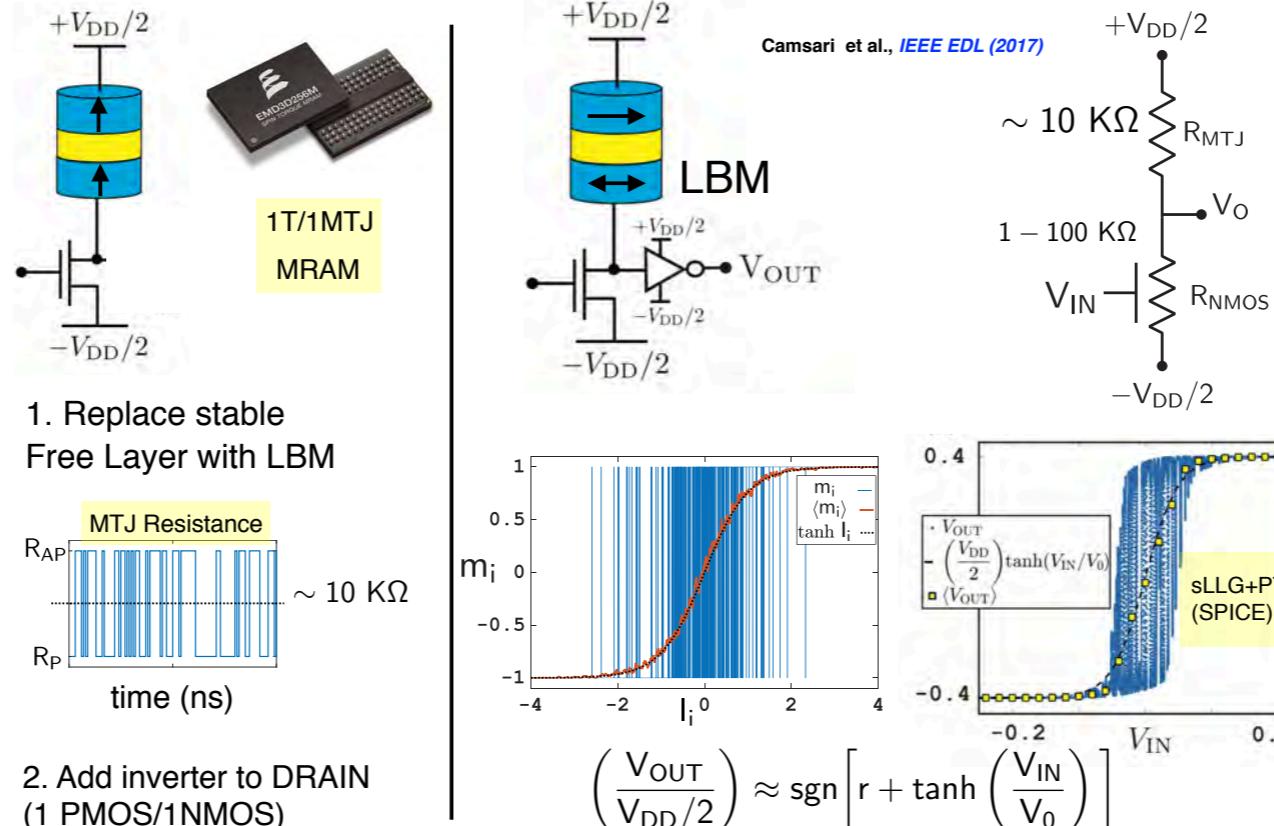
Synapse: Linear weighted sum

*** Update order: RBM qualification

Low-barrier magnets (LBM) are natural p-bits



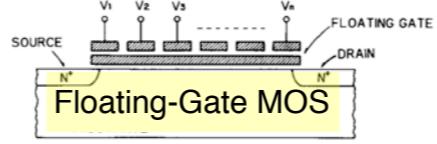
Hardware mapping for p -bit



Possibilities for hardware synapse

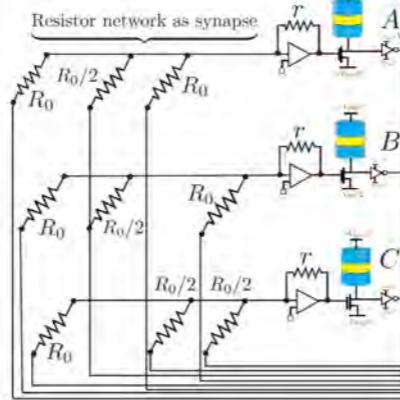
$$m_i = \text{sgn}[\text{rand}(-1, +1) + \tanh(I_i)] \quad (1)$$

$$I_i = I_0 \left(\sum J_{ij} m_j + h_i \right) \quad (2)$$



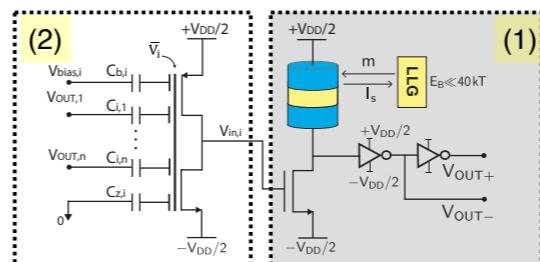
Neu-MOS, Ohmi (1992), IEEE TED

(mem)resistor
cross-bar $[G]_{ij}$



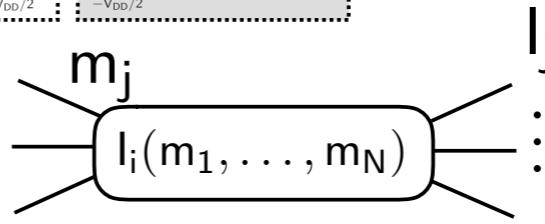
Camsari et al., IEEE EDL (2017)

Capacitive
cross-bar $[C]_{ij}$ Hassan et al., arXiv:1801.0926



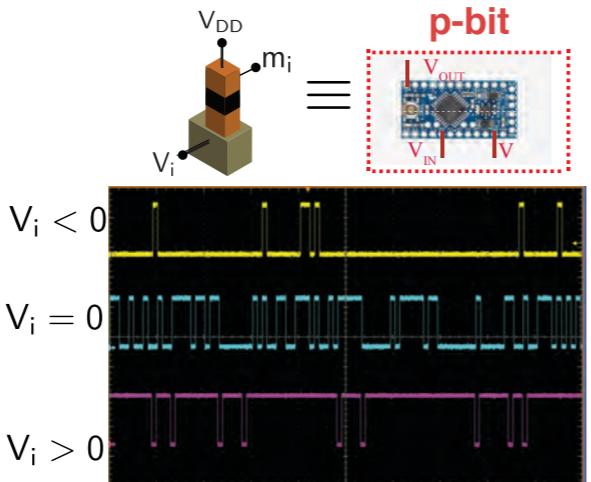
CMOS
circuits

might enable ***non-linear***
synapses, e.g.
 $I_j \propto m_1 m_2 W_{21j}$



Non-magnetic *p*-bits/*p*-circuits

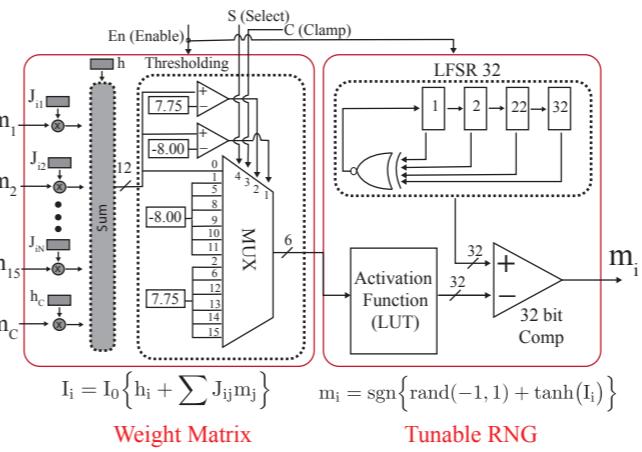
Micro-controller Implementation



Pervaiz et al., *Sci. Rep. (2017)*

- Demonstrated invertible p-circuits up to ~50 p-bits: 4-bit Integer Factorization

FPGA implementation

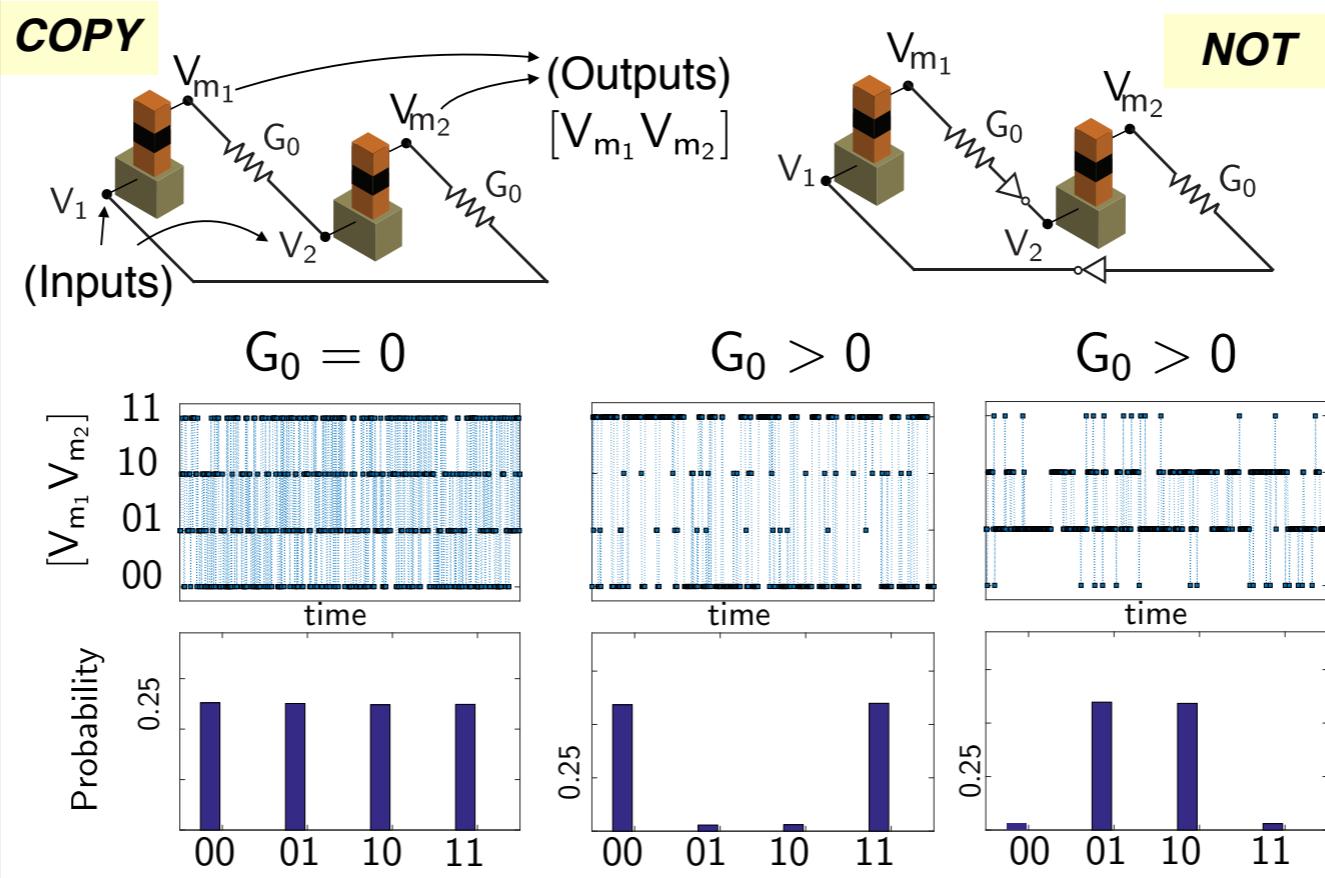


Pervaiz et al., *IEEE TNNLS (2018)*

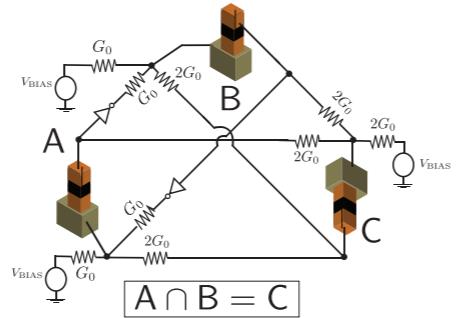
- Demonstrated p-circuits up to ~500 p-bits: Solved a 16 bit Subset Sum Problem

- How to build a p-bit?
- **Quantum Computing-inspired
(Reciprocal Networks)**
- Machine Learning-inspired (Reciprocal
and Non-reciprocal Networks)

Correlated p -bits: p -circuits



Correlated p -bits: p -circuits



$$A \cap B = C$$

$$m_i = \text{sgn}[\text{rand}(-1, +1) + \tanh(I_i)]$$
$$I_i = I_0 \left(\sum J_{ij} m_j + h_i \right)$$

Basic equations

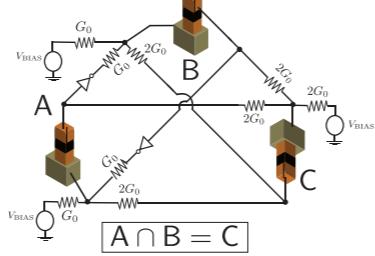
Truth Table

A	B	C
0	0	0
0	1	0
1	0	0
1	1	1

$[J_{\text{AND}}]$

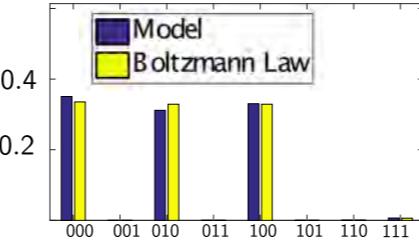
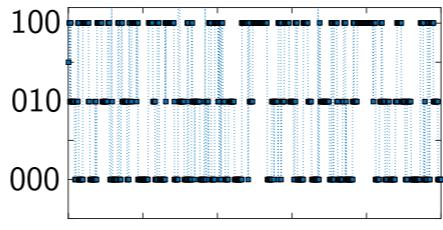
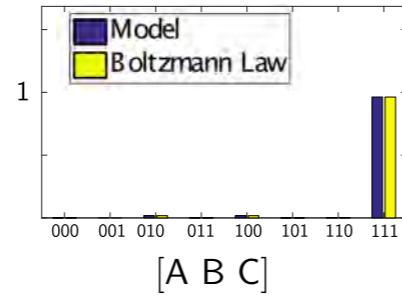
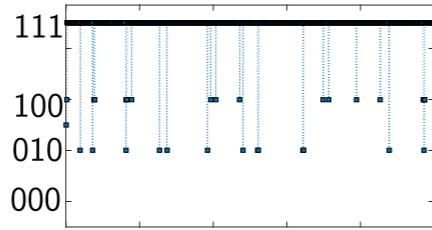
$[G]_{ij}$ network

Invertible Boolean Logic



$$m_i = \text{sgn}[\text{rand}(-1, +1) + \tanh(I_i)]$$
$$I_i = I_0 \left(\sum J_{ij} m_j + h_i \right)$$

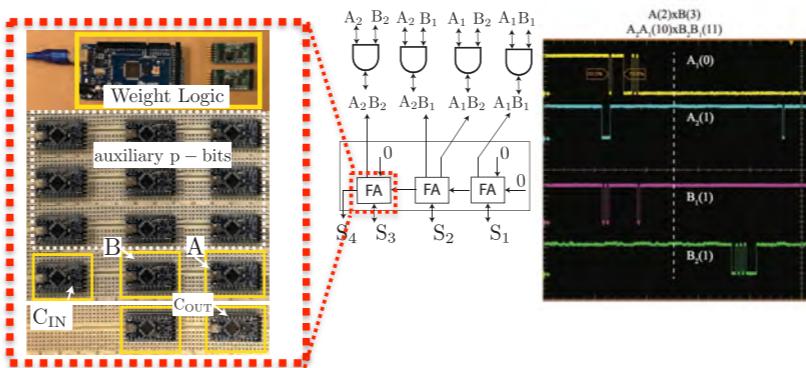
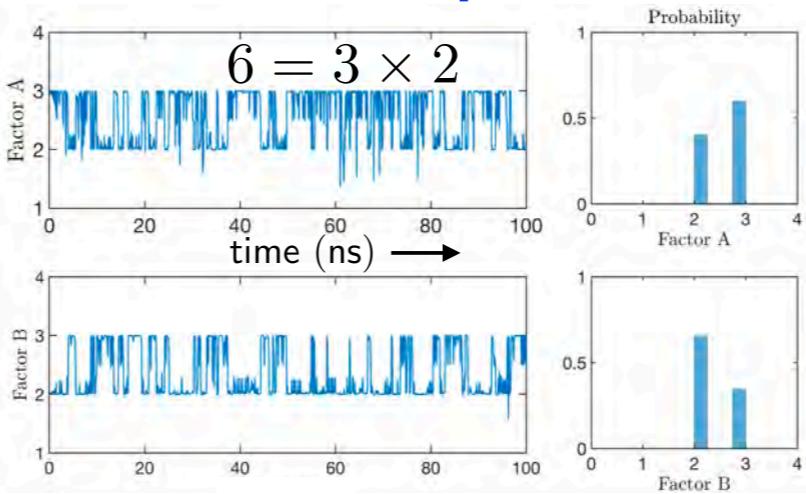
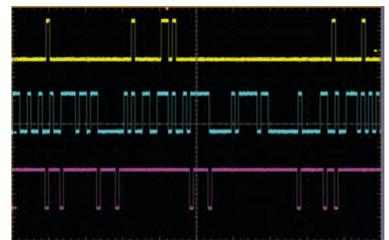
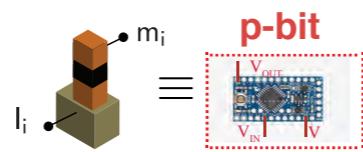
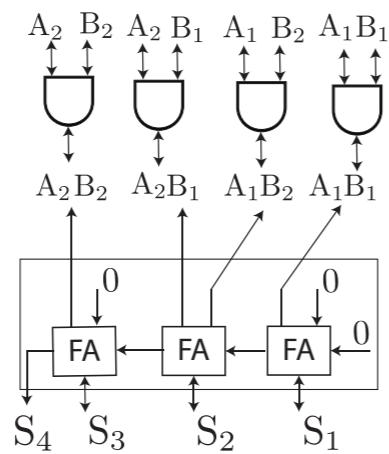
{ h } used as a handle to
“clamp” inputs **and outputs!**



$h_A \rightarrow +1$
 $h_B \rightarrow +1$
 $A \rightarrow 1$
 $B \rightarrow 1$

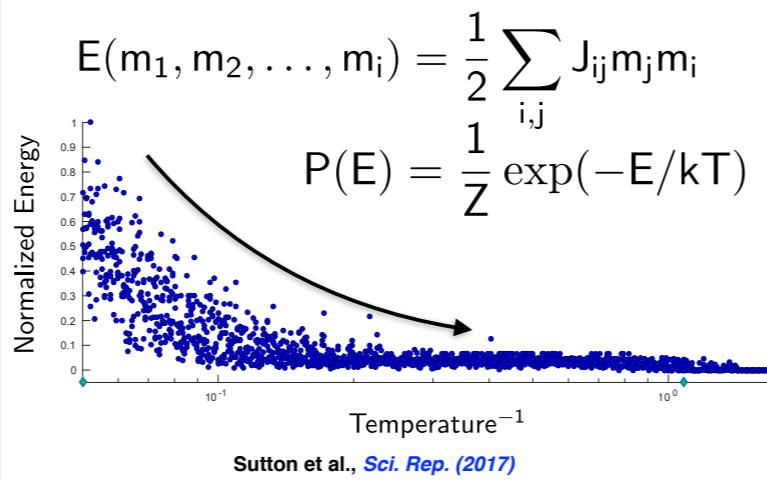
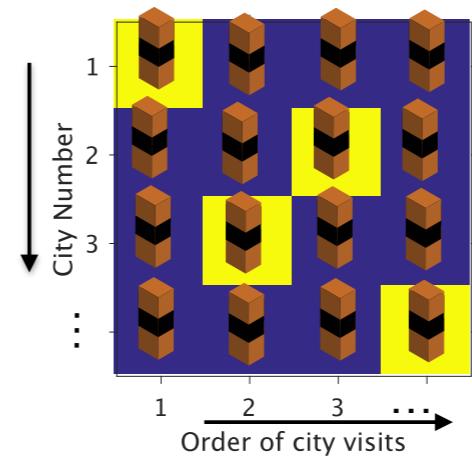
$h_C \rightarrow -1$
 $C \rightarrow 0$

Factorization as Inverse Multiplication



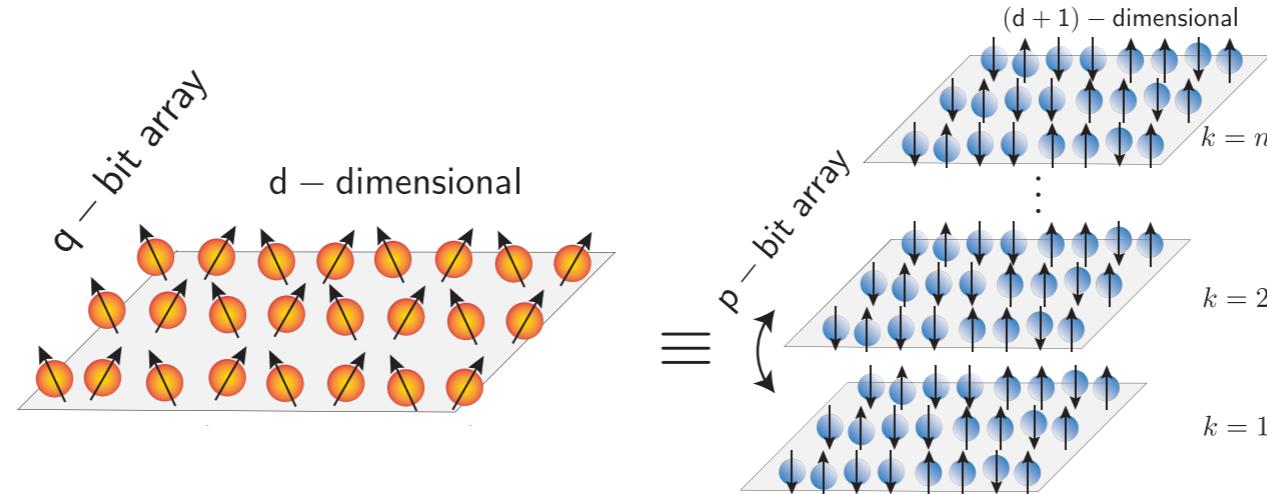
D-Wave reference here

Combinatorial Optimization



- Needs sparse, quantized J
 - Searches the phase space at *GHz rates*
-
- time (ns)

An exact mapping of q-bits to p-bits



$$H_{\text{quantum}} = - \left(\sum_{i < j} J_{i,j} \sigma_i^z \sigma_j^z + \Gamma_x \sum_i \sigma_i^x \right)$$

$$H_{\text{classical}} = - \left(\lim_{n \rightarrow \infty} \sum_{k=1}^n \sum_{i < j} (J_{||})_{i,j} m_{i,k} m_{j,k} + J_{\perp} m_{i,k} m_{i,k+1} \right)$$

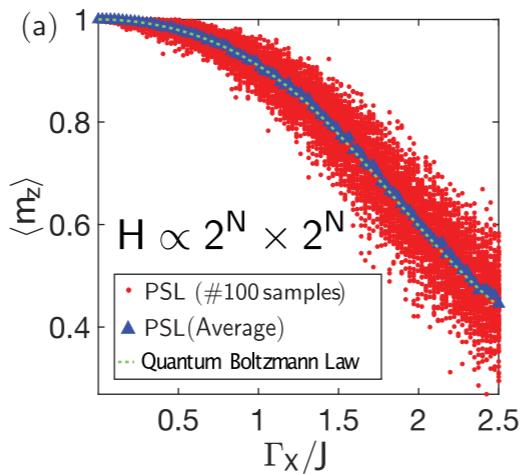
$$J_{||} = J/n \quad J_{\perp} = -\frac{1}{2\beta} \log \tanh(\beta \Gamma/n)$$

- Trotter-Suzuki Decomposition

- Efficient for “stoquastic” Hamiltonians

Quantum Annealing with p-bits

1D spin-chain with M=8 q-bits ($J=+1$)



$$H_{\text{quantum}} = - \left(\sum_{i < j} J_{i,j} \sigma_i^z \sigma_j^z + \Gamma_x \sum_i \sigma_i^x \right)$$

$$\langle m^z \rangle = \frac{\text{Tr.}[m^z \exp(-\beta H)]}{\text{Tr.}[\exp(-\beta H)]}$$

- “Quantum” Boltzmann Law

n = 250 replicas

$$H_{\text{classical}} = - \left(\lim_{n \rightarrow \infty} \sum_{k=1}^n \sum_{i < j} (J_{||})_{i,j} m_{i,k} m_{j,k} + J_{\perp} m_{i,k} m_{i,k+1} \right)$$

- Replicated p-bits vs Exact solution: Averages and *correlations match*

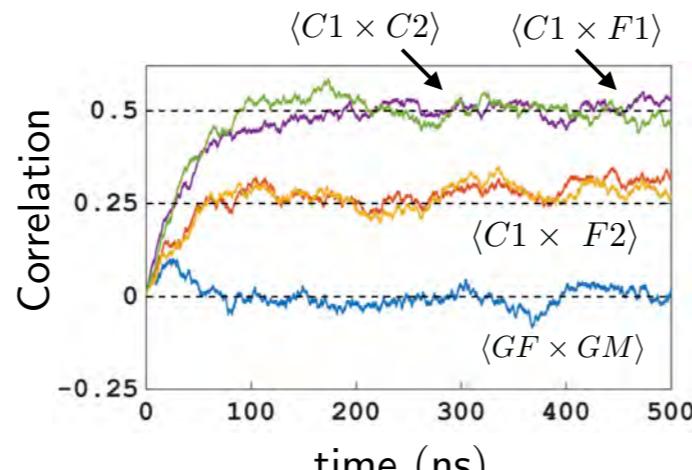
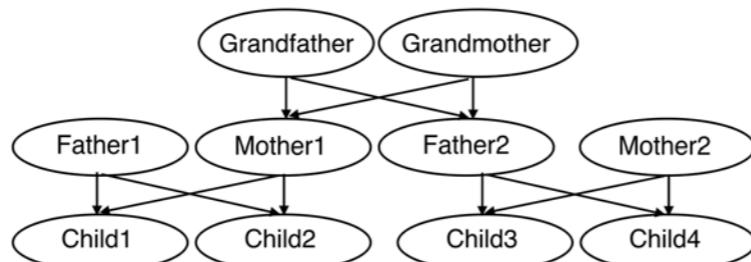
18

Comment: SPICE simulation

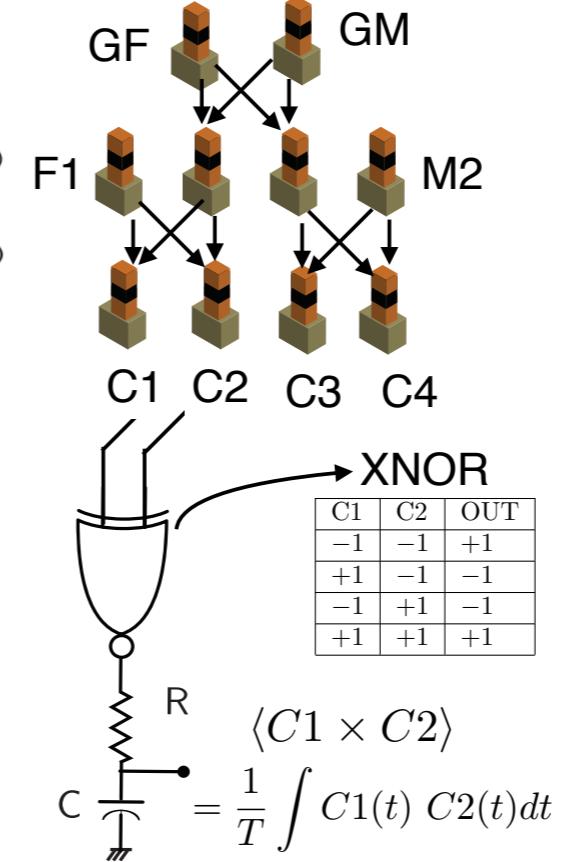
Comment: Size of matrix 2^N by 2^N

- How to build a p-bit?
- Quantum Computing-inspired (Reciprocal Networks)
- **Machine Learning-inspired (Reciprocal and Non-reciprocal Networks)**

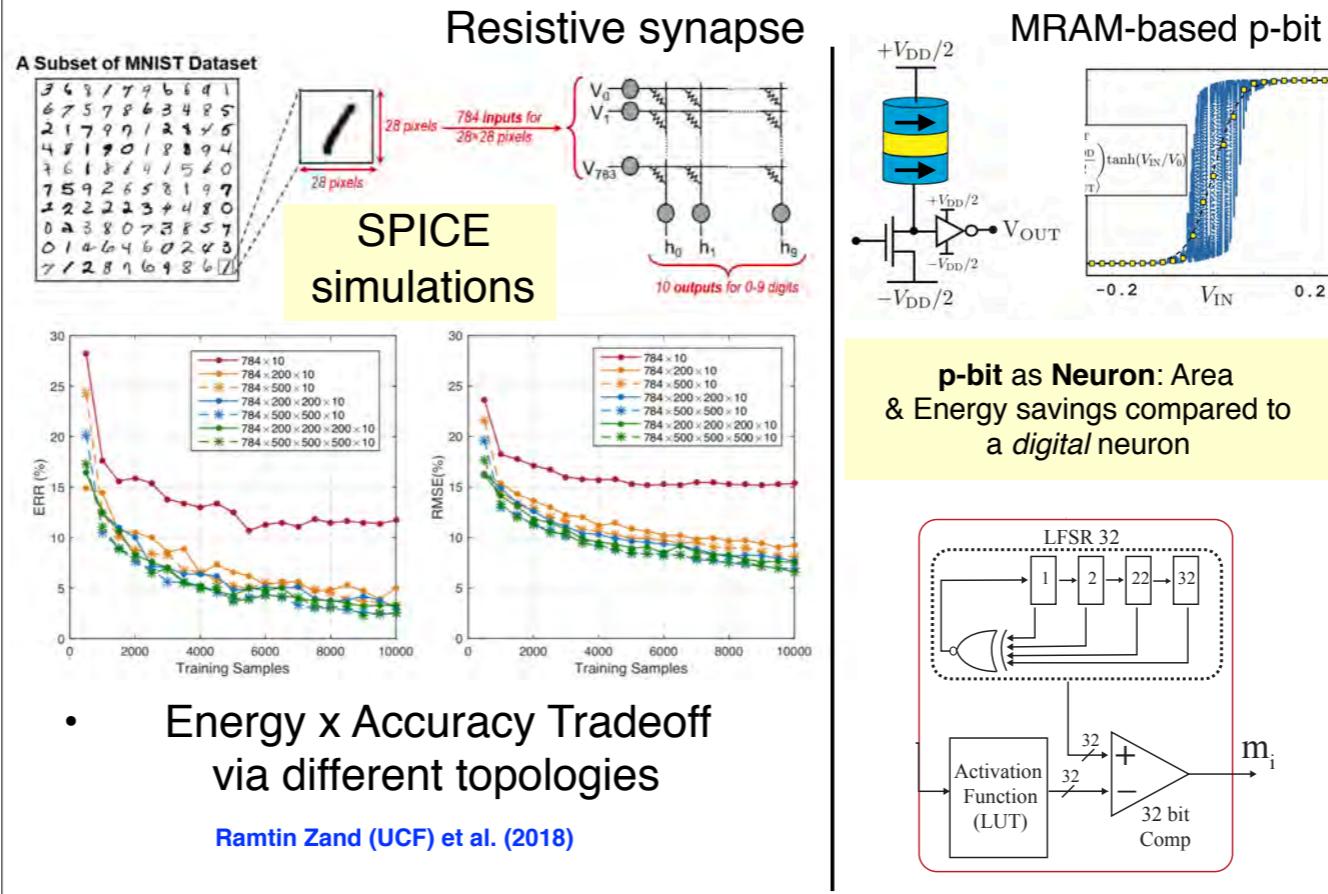
Inherently probabilistic circuits: Bayesian Networks



Faria et al., *AIP Advances* (2018)



Low-power DBN: Inference with p-bits



True North reference: LFSR's are expensive

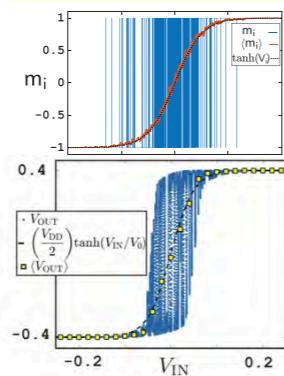
Of course, digital neurons are more reliable and probably offer better scaling / fully thresholded.

Full SPICE could be misleading.

Energy X 1000, could it be defended?

Summary

p-bit as hardware building block



$$m_i = \text{sgn}[\text{rand}(-1, +1) + \tanh(I_i)]$$
$$I_i = I_0 \left(\sum J_{ij} m_j + h_i \right)$$

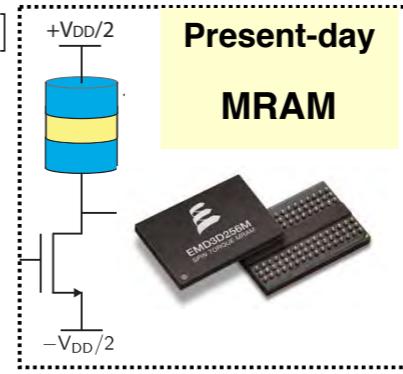
\approx one-to-one mapping

Synapse
[C]_{ij} network
[G]_{ij} network
CMOS

Present-day



MRAM



Quantum Computing

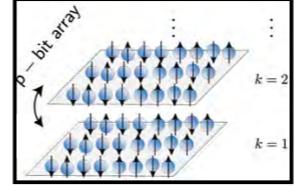
Invertible Logic

Adder \leftrightarrow Subtractor
(1) Subset Sum Problem
Multiplier \leftrightarrow Divider
(2) Prime Factorization

Optimization

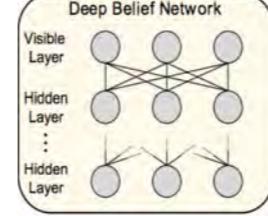


Quantum Circuits



Machine Learning

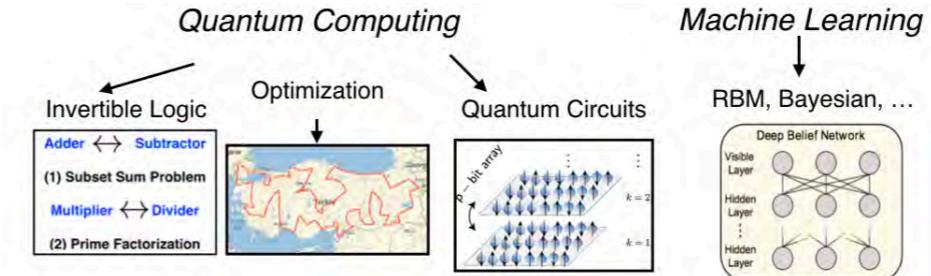
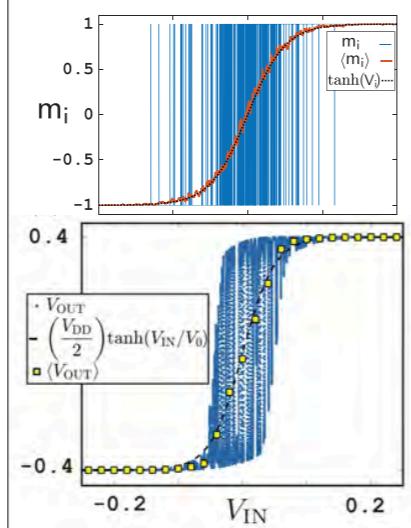
RBM, Bayesian, ...



p-bits for PSL

$$m_i = \text{sgn}[\text{rand}(-1, +1) + \tanh(l_i)]$$

$$l_i = l_0 \left(\sum J_{ij} m_j + h_i \right)$$



The End of Moore's Law:
A New Beginning for Information Technology



T. N. Theis and
P. Wong, 2016

Applications/Algorithms: Quantum Computing / Machine Learning

***p*-bits & *p*-circuits as**

Novel devices and circuits for “*the algorithms and applications of the coming decades*”

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