$\label{eq:multi-SPU} \mbox{Multi-GPU accelerated multi-spin Monte Carlo simulations} \\ \mbox{of the 2D Ising model}$



Ising model

$$H = -J \sum_{\langle i,j \rangle} S_i S_j$$

Metropolis algorithm:

$$p_a = e^{-\frac{E}{k_B T}}$$

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Mutliple GPUs



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How does it scale? Is it worth the effort?

► Formula:

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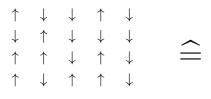
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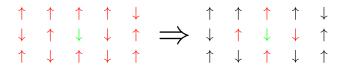
► System sizes: 100′000 × 100′000

Nearest neighbour interactions only!

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Calculation of energy: $\mathcal{O}\left(n^2\right) \Rightarrow \mathcal{O}\left(n\right)$

► Goal: Sample phase space



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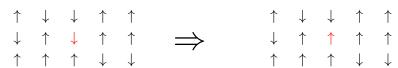
Algorithm:

► Goal: Sample phase space



Algorithm:

1.) Propose new state: Random spin flips!



► Goal: Sample phase space



Algorithm:

1.) Propose new state: Random spin flips!

2.) Accept the new configuration: $p_a = e^{-\frac{\Delta E}{k_B T}}$

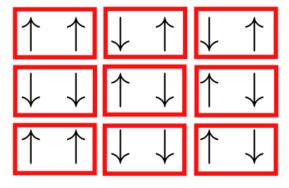
Single core CPU: Data structure

lacksquare Multi-spin coding: 1 spin $\widehat{=}$ 1 bit

.

Single core CPU: Data structure

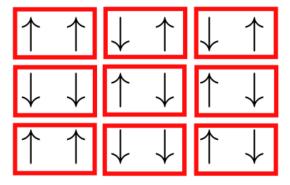
- ▶ Multi-spin coding: $1 \text{ spin } \widehat{=} 1 \text{ bit}$
- ► Group spins in groups of size 32 (int)



Ę

Single core CPU: Data structure

- ▶ Multi-spin coding: $1 \text{ spin } \widehat{=} 1 \text{ bit}$
- Group spins in groups of size 32 (int)



▶ $100'000 \times 100'000$ lattice: ≈ 1.2 GB

Type:	# opposed	ΔE caused by flip of \uparrow
↑ ↑ ↑ ↑	0	+8J
↑ ↑ ↑ ↑	1	+4J
† † † †	2	0
→ → → →	3	-4J
→ → → →	4	-8J

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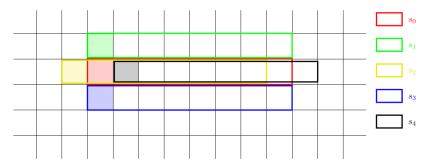
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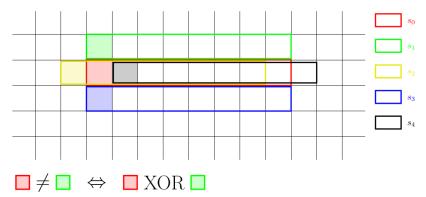
Type:	# opposed	ΔE caused by flip of \uparrow
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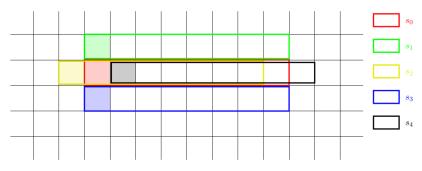
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Count #opposed spins!







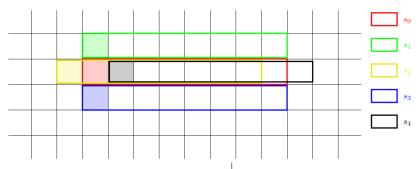
$$\square \neq \square \Leftrightarrow \square XOR \square$$

$$i_1 = \square XOR \square$$

 $i_2 = \square XOR \square$

$$i_3 = \square XOR \square$$

$$i_4 = \square XOR \square$$



$$\square \neq \square \Leftrightarrow \square \text{ XOR } \square$$
 $i_1 = \square \text{ XOR } \square$

$$i_2 = \square XOR \square$$
 $i_2 = \square XOR \square$

$$i_3 = \square XOR \square$$
 $i_4 = \square XOR \square$

Combine with acceptance probability:

$$i_1+i_2+i_3+i_4+2\exp_8+\exp_4 \ge 2$$

► Port CPU implementation

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 - © Bad performance!

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Shared memory:



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Single GPU implementation I

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Shared memory:



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Reduce # accesses to global memory!

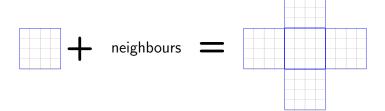
Single GPU implementation II

- ► Metaspins (4 × 4)
 - $\rightarrow 1$ metaspin $\ensuremath{\widehat{=}}\xspace 2$ bytes = 1 unsigned short int
 - ightarrow only 1 read per metaspin

9

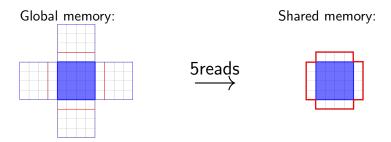
Single GPU implementation II

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 - ightarrow 1 metaspin $\widehat{=}$ 2 bytes = 1 unsigned short int
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- ► Read:



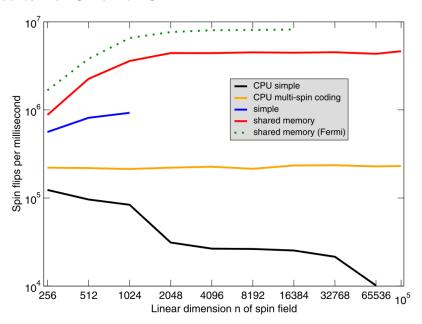
9

Single GPU implementation II



 \Rightarrow 5 reads to flip entire metaspin!

Results: CPU vs. GPU



► Single GPU: fast

► Single GPU:

```
fast system size \leq 4 GB ( \widehat{=}\ 100'000\times 100'000)
```

Single GPU:
 fast
 system size ≤ 4 GB (= 100′000 × 100′000)
 Idea: Distributed lattice!

► Single GPU: fast system size \leq 4 GB ($\widehat{=}$ 100′000 \times 100′000)

- ▶ Idea: Distributed lattice!
- ► Algorithm:
 - 1. Copy neighbour borders to GPU

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```
fast system size \leq 4 GB (\hat{=} 100′000 \times 100′000)
```

- Idea: Distributed lattice!
- Algorithm:
 - 1. Copy neighbour borders to GPU
 - 2. Update own region on GPU
 - 3. Copy boundary spins to CPU
 - 4. Exchange boundary spins with other nodes

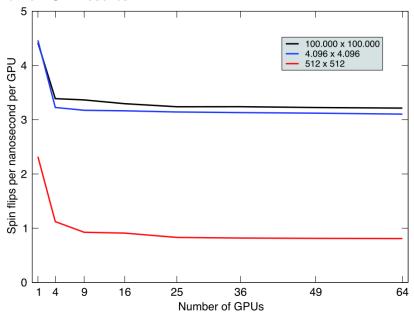
► Single GPU:

fast

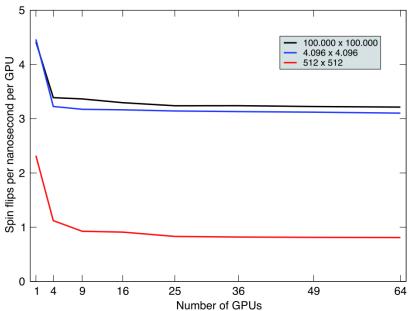
system size < 4 GB (= 100′000 × 100′000)

- Idea: Distributed lattice!
- Algorithm:
 - 1. Copy neighbour borders to GPU
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 - 4. Exchange boundary spins with other nodes
 - 5. Repeat or finish

Multi-GPU: Results



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64 GPUs, $800'000 \times 800'000$: 3s

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Why this paper?!