

# QMC of everything: A universal algorithm for simulating arbitrary quantum many-body systems

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Based partly on: [arXiv:2307.06503](https://arxiv.org/abs/2307.06503), [arXiv:2006.02539](https://arxiv.org/abs/2006.02539)

IPAM workshop: Many-body Quantum Systems  
via Classical and Quantum Computation

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Information Sciences Institute  
USC School of Engineering

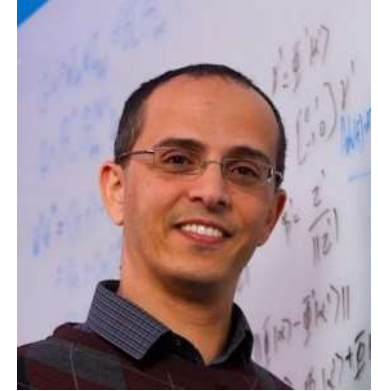
# Main collaborators and funding



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## □ Funding:



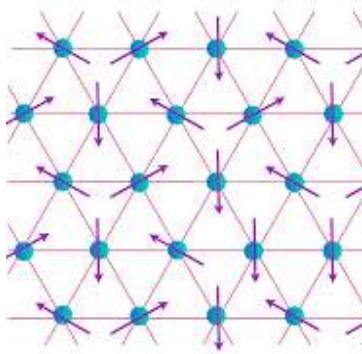
# Outline

- A brief intro to quantum Monte Carlo (QMC) simulations
- Permutation Matrix Representation QMC
- QMC of everything: Provably ergodic Markov chains for arbitrary spin  $1/2$  systems
- Generalization to higher-spin particles, fermions and bosons
- PMR in quantum computing
- Summary and conclusions

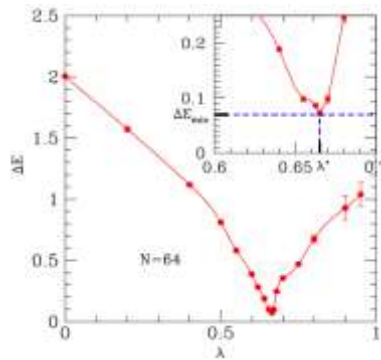
# A brief intro to QMC simulations

# Motivation

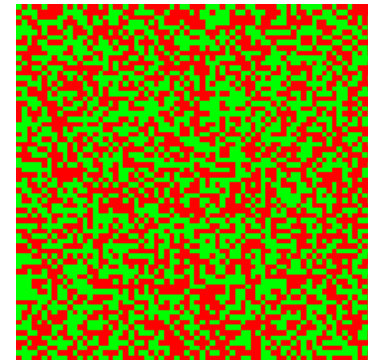
- Generally, large-scale quantum many-body systems cannot be studied analytically. Too many degrees of freedom.
- Quantum Monte Carlo (QMC) sampling is often the only approach available to study these without approximations.
- Still, **QMC is known to be inefficient in very many cases.**



*sign problem  
(negative weights)*



*small energy gaps  
(phase transitions)*



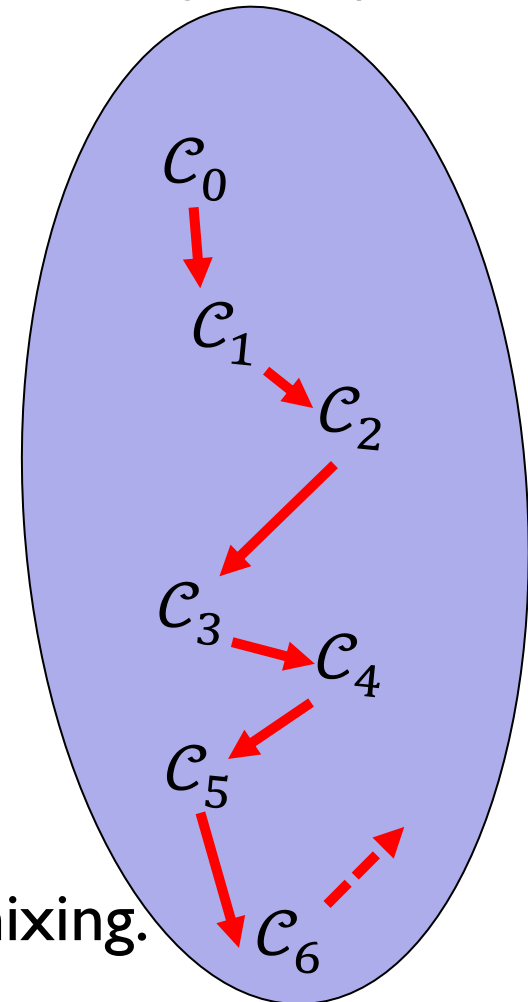
*classical glassiness*

- **Also: for each model, one usually needs a specially tailored algorithm.**

# What is QMC?

- Goal is to write  $Z = \text{Tr}[e^{-\beta H}]$  as a sum of *non-negative* weights  $W_c$  for configurations  $\mathcal{C}$ :  $Z = \sum_c W_c$ .
- Similarly write observables as  $\langle A \rangle = \frac{1}{Z} \text{Tr}[A e^{-\beta H}]$  as  $\frac{1}{Z} \sum_c A_c W_c$ .
- Use Markov chain to sample each  $\mathcal{C}$  with probability  $p_c = \frac{W_c}{Z}$ .
- **Collect statistics about  $A$**  by measuring  $A_c$ . Visit important configurations more often.
- Hope for (I) no sign problem and (II) rapid mixing.

Markov Chain Monte Carlo (MCMC)  
in QMC configuration space



# Standard QMC approaches

- Path integral Monte Carlo methods are prone to **Trotterization errors**.

$$e^{-\beta H} = \left(e^{-\Delta t H}\right)^L \approx \left(e^{-\Delta t A} e^{-\Delta t B}\right)^L \quad \text{but} \quad e^{\Delta t A} e^{\Delta t B} \neq e^{\Delta t (A+B)}$$

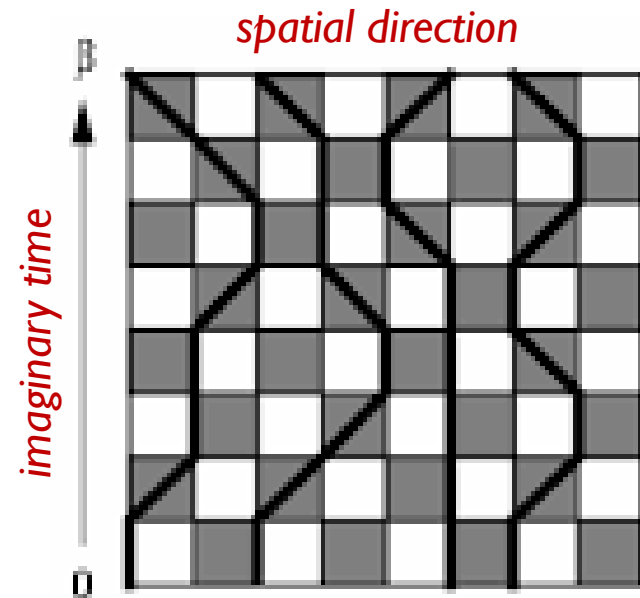
- Other schemes which are immune to Trotterization errors:

- Continuous-time path integral Monte Carlo  
[pioneered by Prokof'ev et al].

- Stochastic series expansion (SSE)  
[pioneered by Sandvik].

- These have their own “issues”.

- **Common annoyance: need to design update rules on a model-by-model basis.**

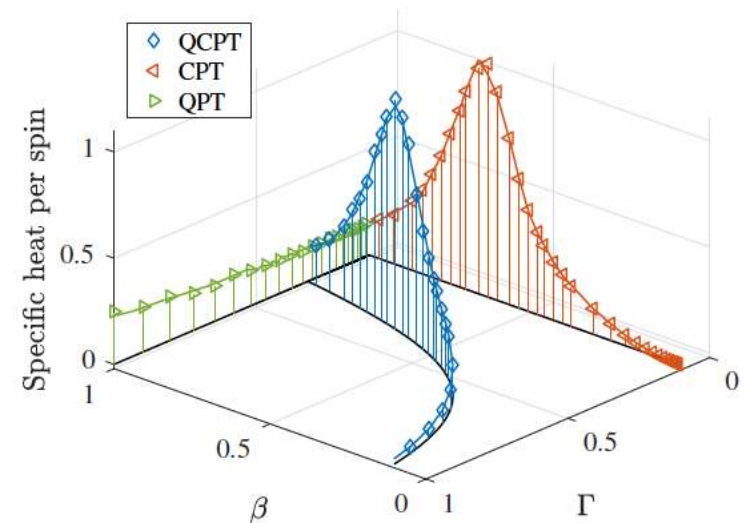


# Permutation Matrix Representation Quantum Monte Carlo



# Permutation matrix representation

- ❑ **PMR-QMC** is a novel series expansion of the quantum partition function (T. Albash, G. Wagenbreth, IH, PRE, 2017).
- ❑ It is *parameter-free* and *Trotter-error free*.
- ❑ It is *universal* and *abstract*: it applies essentially to any model and does not require model-specific adjustments.
- ❑ It does *not* solve the sign problem but can be used to solve or mitigate it in certain cases.
- ❑ Becomes a thermal classical simulation in the classical limit.

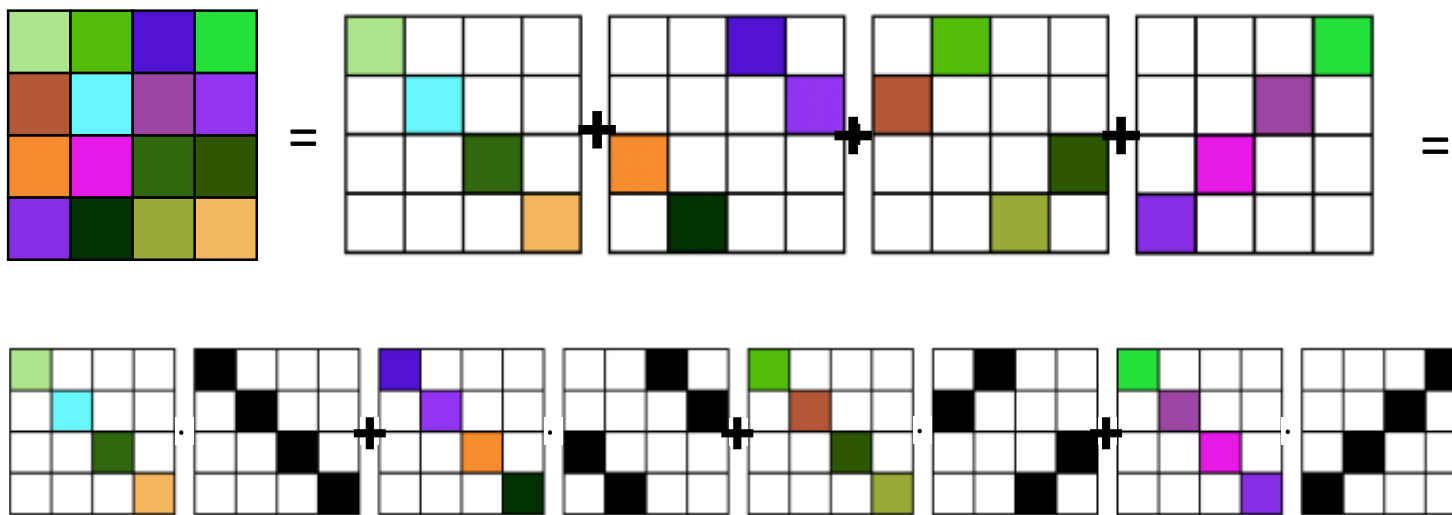


# Permutation matrix representation

- Choose a basis  $\{|z\rangle\}$  (call it the “computational” basis).
- Cast Hamiltonian as a sum of  $M$  monomial (generalized permutation) operators  $\tilde{P}_j = D_j P_j$  (always possible and easy to do):

$$H = \sum_{j=0}^M \tilde{P}_j = \sum_{j=0}^M D_j P_j = D_0 + \sum_{j=1}^M D_j P_j$$

Permutation matrix representation of the Hamiltonian



# Permutation matrix representation

$$H = \sum_{j=0}^M \tilde{P}_j = \sum_{j=0}^M D_j P_j = D_0 + \sum_{j=1}^M D_j P_j$$

- Sum of **products of diagonal (non-local) operators  $D_j$  and permutation operators  $P_j$** , where  $P_j |z\rangle = |z'\rangle (\neq |z\rangle)$ .
- $P_0 = \mathbb{I}$  making  $D_0$ , the “classical”/diagonal Hamiltonian.
- Rest of permutations  $P_j$  have *no fixed points*.
- E.g., for spin models,  $P_j$ 's would be  $\mathbb{I}, X_1, X_2, \dots, X_1 X_2, \dots$
- The  $D_j$ 's are generally non-local diagonal operators.

# The off-diagonal series expansion

- Expand the partition function  $Z = \text{Tr}[e^{-\beta H}]$  in a Taylor series substituting  $\text{Tr}[\cdot] = \sum \langle z | \cdot | z \rangle$ :

$$Z = \sum \left\langle z \left| \sum \frac{(-\beta)^n}{n!} \left( D_0 + \sum D_j P_j \right)^n \right| z \right\rangle.$$

- Expanding, we get

$$Z = \sum \sum \frac{(-\beta)^n}{n!} \left\langle z \left| \begin{array}{l} \text{sum of all possible products} \\ \text{of length } n \text{ of } D_0 \text{ and } D_j P_j \end{array} \right| z \right\rangle.$$

T.Albash, G.Wagenbreth and I. Hen, “Off-Diagonal Expansion Quantum Monte Carlo”, Phys. Rev. E **96**, 063309 (2017).

# The off-diagonal series expansion

□ Take one such product for example:

$$\langle z | D_0 \cdot D_0 \cdot D_1 P_1 \cdot D_0 \cdot D_1 P_1 \cdot D_0 \cdot D_0 \cdot D_2 P_2 \cdot D_0 | z \rangle$$

$$= \langle z | D_0 \cdot D_0 \cdot D_1 P_1 \cdot D_0 \cdot D_1 P_1 \cdot D_0 \cdot D_0 \cdot D_2 P_2 | z \rangle \cdot E(z)$$

$$= \langle z | D_0 \cdot D_0 \cdot D_1 P_1 \cdot D_0 \cdot D_1 P_1 \cdot D_0 \cdot D_0 P_2 | z \rangle \cdot d_2(z') E(z)$$

$$= \langle z | D_0 \cdot D_0 \cdot D_1 P_1 \cdot H_c \cdot D_1 P_1 P_2 | z \rangle \cdot E^2(z') d_2(z') E(z)$$

$$\vdots$$

$$\langle z | P_1 P_1 P_2 | z \rangle \cdot E^2(z''') d_1(z''') E(z'') d_1(z'') E^2(z') d_2(z') E(z)$$

→ This is the “off-diagonal core” consisting of a unique product of permutation matrices. Evaluates either to 0 or to 1.

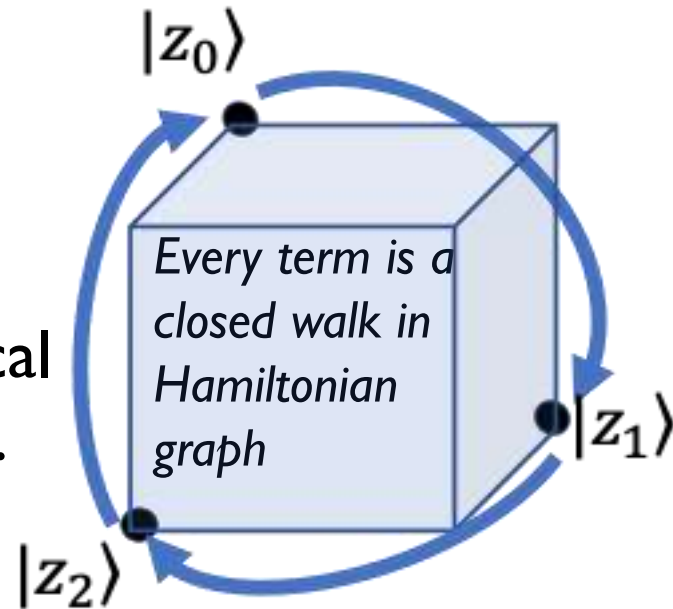
# The off-diagonal series expansion

- There is an infinite number of terms that share the same off-diagonal core  $\langle z | P_{i_q} \dots P_{i_2} P_{i_1} | z \rangle$ . We group them together.
- Each pair  $\{ |z\rangle, S_{\vec{i}_q} = P_{i_q} \dots P_{i_2} P_{i_1} \}$  defines a closed walk on the Hamiltonian graph (where basis states are nodes, off-diagonal elements are edges):

$$|z\rangle = |z_0\rangle \xrightarrow{P_{i_1}} |z_1\rangle \rightarrow \dots \rightarrow |z_q\rangle = |z\rangle$$

- This path "induces" a sequence of classical energies  $D_0 |z\rangle = E_z |z\rangle: E_{z_0}, E_{z_1}, \dots, E_{z_q}$ .

- And accompanying  $d_{i_1}(z_1), d_{i_2}(z_2), \dots$



# Grouping with divided differences

- Summing the contributions of all terms that share the same off-diagonal core, we get the *infinite sum*

$$\left( \sum_{n=q}^{\infty} \frac{\beta^n (-1)^{n-q}}{n!} \sum_{\sum k_i = n-q} E_c^{k_0}(z_0) \cdot \dots \cdot E_c^{k_q}(z_q) \right)$$

- As it turns out, this sum can be neatly evaluated to the easily computable

$$e^{-\beta [E_{z_0}, E_{z_1}, \dots, E_{z_q}]}$$

This is the *divided difference* of the Boltzmann factors of sequences (multi-sets) of intermediate classical energies along the imaginary time direction.

# Digression: divided differences



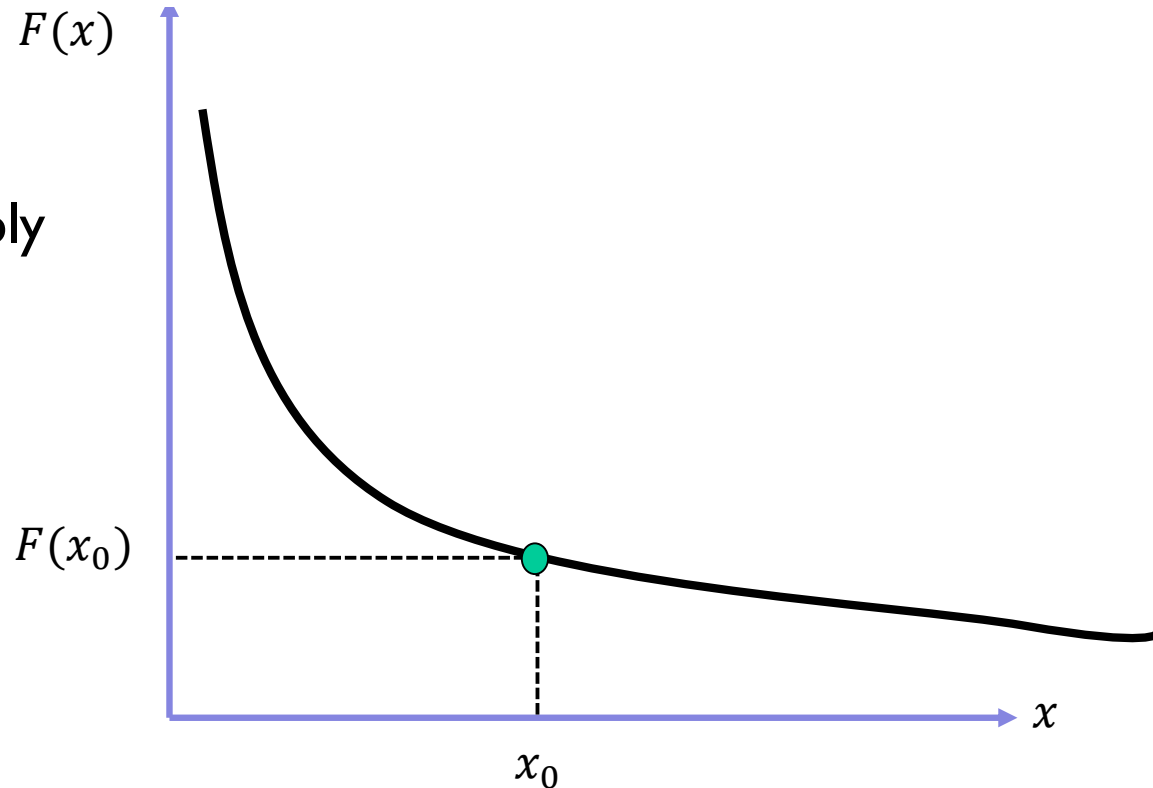
# Digression: divided differences

- The divided differences of a function  $F(x)$  with respect to the input multi-set  $[x_0, \dots, x_q]$  is given by:

$$F([x_0, \dots, x_q]) \equiv \sum_{j=0}^q \frac{F(x_j)}{\prod_{k \neq j} (x_j - x_k)}$$

- The divided difference of a function with an input multi-set of size one, is simply

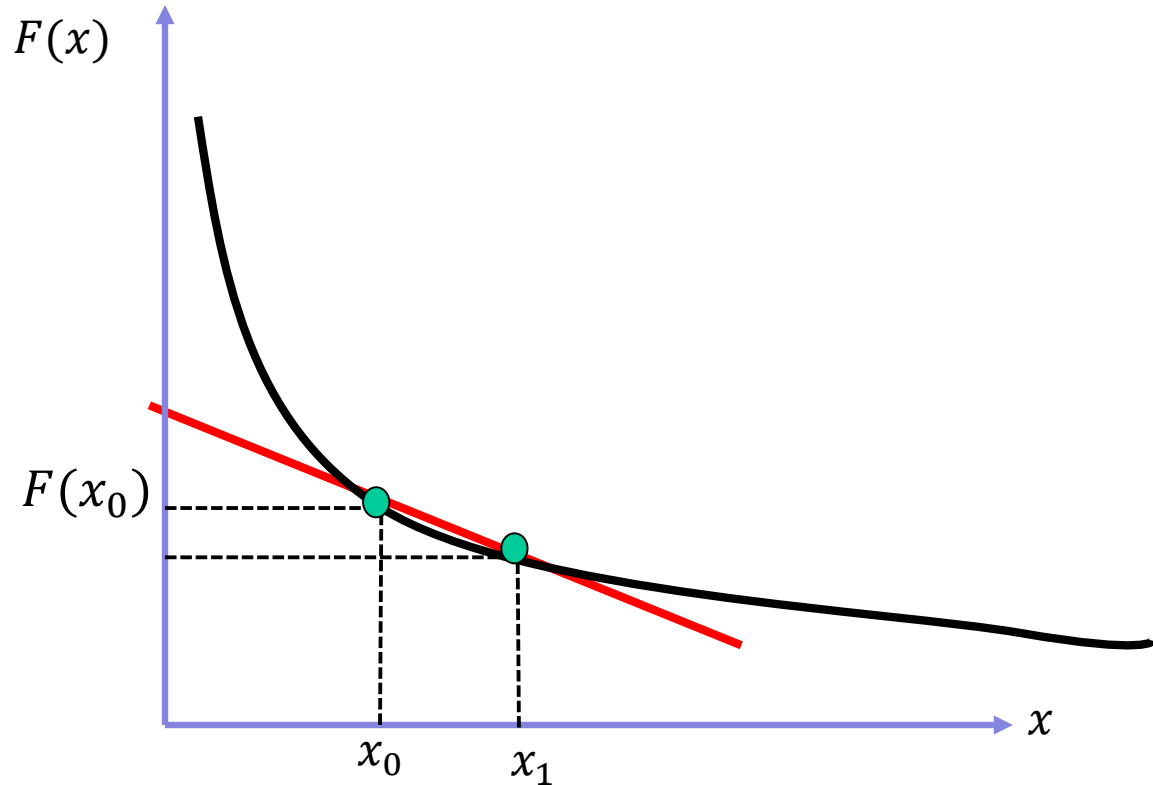
$$F[x_0] = F(x_0)$$



# Digression: divided differences

- The divided differences of a function taking as input a multi set with two elements is:

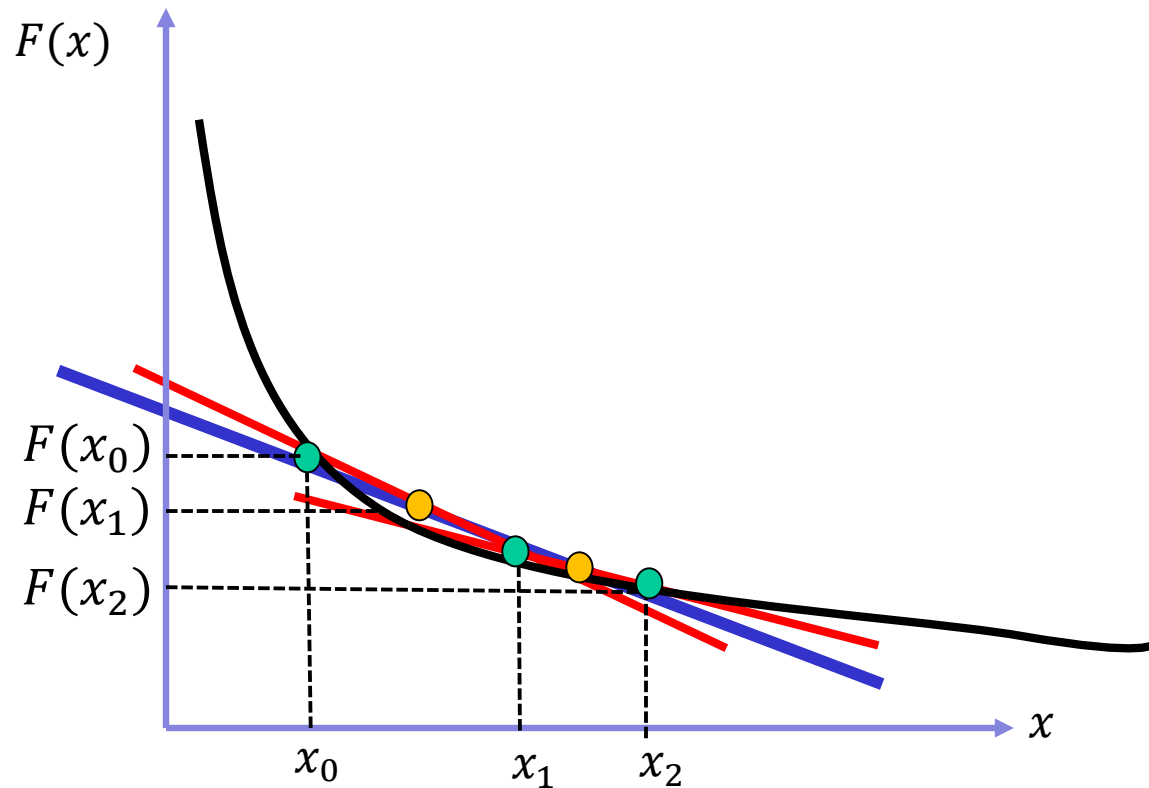
$$F[x_0, x_1] = \frac{F(x_1) - F(x_0)}{x_1 - x_0} \approx F'(\xi)$$



# Digression: divided differences

- The divided differences of a function taking as input a multi set with three elements is:

$$F[x_0, x_1, x_2] = \frac{F[x_0, x_1] - F[x_1, x_2]}{x_0 - x_2} \approx \frac{1}{2} F''(\xi)$$



# Digression: divided differences

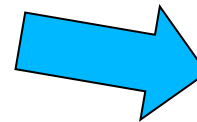
- In the general case, the evaluation of the divided differences of a function with  $q + 1$  inputs

$$F([x_0, \dots, x_q]) \equiv \sum_{j=0}^q \frac{F(x_j)}{\prod_{k \neq j} (x_j - x_k)}$$

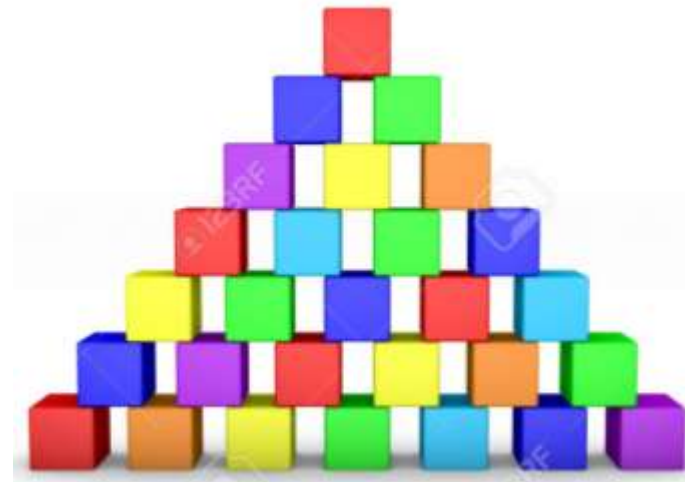
is done via the recursion relation:

$$F[x_0, \dots, x_q] = \frac{F[x_0, \dots, x_{q-1}] - F[x_1, \dots, x_q]}{x_0 - x_q}$$

- Also: 
$$F[x_0, \dots, x_q] = \frac{F^{(n)}(\xi)}{n!}$$



- The computational cost of calculating divided differences scales as  $q^2$  in the worst case.



# The partition function

- Grouping all terms together, we end up with the final form

$$Z = \sum_{\substack{z, S_{\vec{i}_q} \\ = \mathbb{I}}} D_{(z, S_{\vec{i}_q})} e^{-\beta [E_{z_0}, E_{z_1}, \dots, E_{z_q}]}$$

$Z$  is a sum over all basis states  $\{|z\rangle\}$  and all the combinations of products of  $P_j$  operators

$D_{(z, S_{\vec{i}_q})}$  is a product of  $q$  off-diagonal matrix elements

This is the divided-difference exponential whose inputs are the classical energies along the walk. Its sign is  $(-1)^q$ .

$S_{\vec{i}_q} = P_{i_q} \dots P_{i_2} P_{i_1}$  (with  $q$  the size of the sequence) that evaluate to  $S_{\vec{i}_q} = \mathbb{I}$ .

□ Each summand is a “generalized

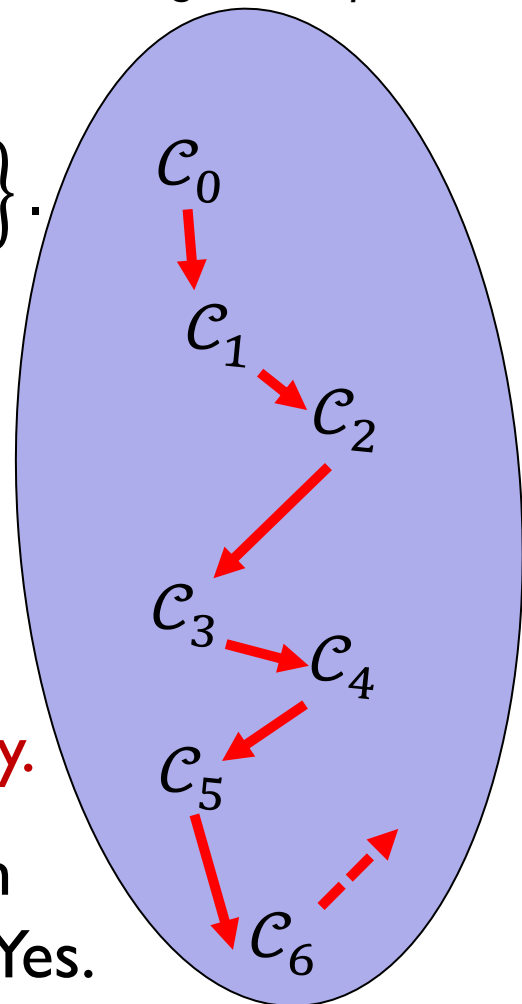
Boltzmann weight”  $W = D_{(z, S_q)} e^{-\beta [E_{z_0}, E_{z_1}, \dots, E_{z_q}]}$ .

# Provably ergodic QMC for arbitrary spin $\frac{1}{2}$ systems

# Guaranteeing ergodicity

- Recall we have a Markov chain to sample each  $\mathcal{C}$  with probability  $p_{\mathcal{C}} = \frac{W_{\mathcal{C}}}{Z}$ .
- Here,  $\mathcal{C}$  is a pair  $\{ |z\rangle, S_{\vec{i}_q} = P_{i_q} \dots P_{i_2} P_{i_1} (= \mathbb{I}) \}$ .
- Covering the  $|z\rangle$  space is easy – “classical moves”, e.g., spin flips.
- **We must also make sure the process is ergodic for any set of  $\{P_i\}$  appearing in  $H$ .**
- **This becomes an interesting problem in theory.**
- Can we produce a finite set of moves that can generate all sequences  $S_{\vec{i}_q}$  that evaluate to  $\mathbb{I}$ ? Yes.

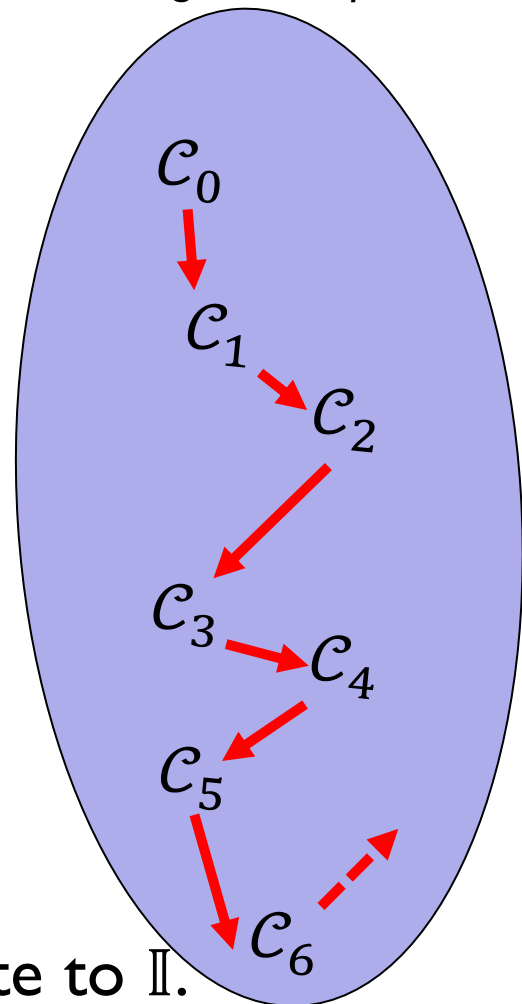
*Markov Chain Monte Carlo (MCMC)  
in QMC configuration space*



# Guaranteeing ergodicity

- Take for example, spin  $1/2$  systems. Here the  $\{P_i\}$  are Pauli- $X$  strings.
- As group elements, they obey  $P_i = P_i^{-1}, [P_i, P_j] = 0$ .
- We were able to prove that all  $S_{\vec{i}_q}$  can be generated with:
  - Operator swapping, e.g.,  $P_3P_1P_2P_3 \rightarrow P_3P_2P_1P_3$
  - Pair insertion/deletion, e.g.,  $P_2P_1 \rightarrow P_2P_3P_3P_1$
  - Insertion and deletion of the “**fundamental cycles**” of the model, e.g.,  $P_1P_2 \rightarrow P_2P_4P_5P_6P_1$
- These are products of distinct  $P_i$ 's that evaluate to  $\mathbb{I}$ .

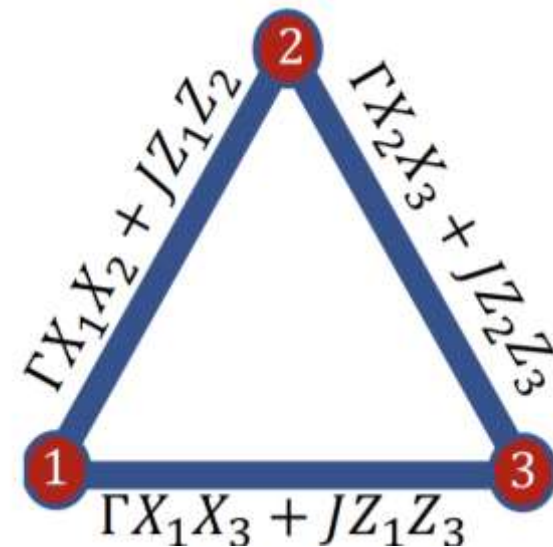
Markov Chain Monte Carlo (MCMC)  
in QMC configuration space





# Guaranteeing ergodicity

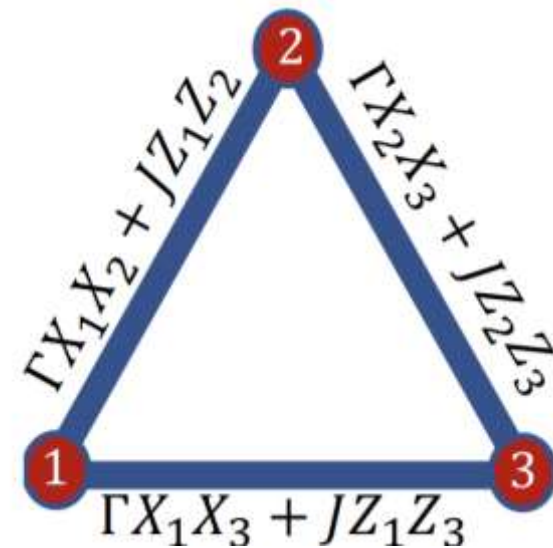
- On a triangular lattice, we may have:  
 $P_1 = X_2X_3$ ,  $P_2 = X_1X_3$  and  $P_3 = X_1X_2$ .
- Here a fundamental cycle is  $P_1P_2P_3 = \mathbb{I}$ .
- Why? Denote:  
 $P_1 \equiv [0,1,1]$ ,  $P_2 \equiv [1,0,1]$ ,  $P_3 \equiv [1,1,0]$ .
- Products of permutations correspond to mod-2 addition of these vectors.
- The fundamental cycles corresponds to the null-space vectors of the  $P$  matrices:  $[0,1,1] \oplus [1,0,1] \oplus [1,1,0] = [0,0,0]$ .
- **All fundamental cycles may be found automatically and efficiently!**



*XY model on the triangular lattice.*

# Guaranteeing ergodicity

- On a triangular lattice, we may have:  
 $P_1 = X_2X_3$ ,  $P_2 = X_1X_3$  and  $P_3 = X_1X_2$ .
- Here a fundamental cycle is  $P_1P_2P_3 = \mathbb{I}$ .
- E.g., the sequence  $P_2P_1P_2P_1P_3P_1P_2$ .
- Via local swaps:  $P_1P_1P_1P_2P_2P_2P_3$ .
- Pair deletions:  $P_1P_2P_3$ .
- Fundamental cycle deletion:  $\mathbb{I}$ .
- This leads to update moves with verified ergodicity for any and all spin  $1/2$  systems!



*XY model on the triangular lattice.*

# Generalization to higher-spin particles, fermions and bosons

# Other types of systems?

## □ What about fermions?

Can be mapped via a Jordan-Wigner transformation to spin  $1/2$  particles.

## □ What about higher spin particles?

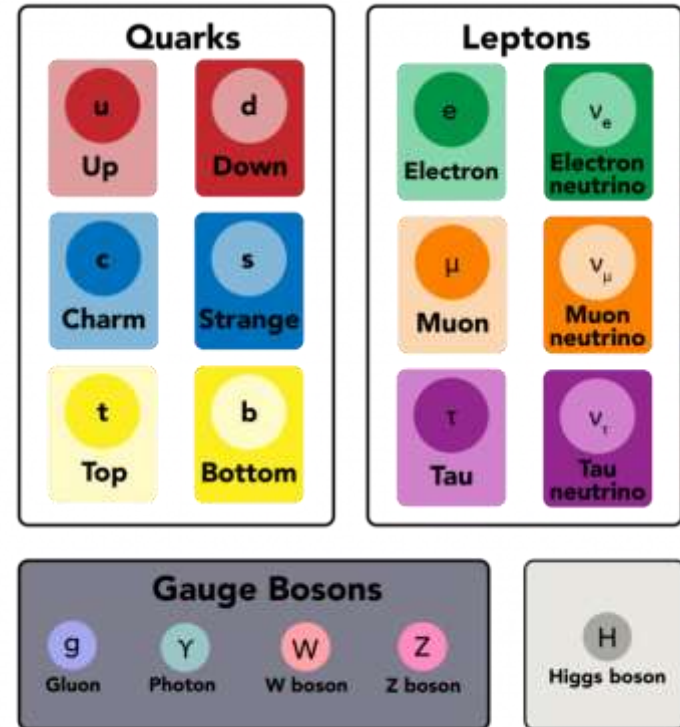
Slightly less trivial group structure:

$P_i = P_i^{-1} \rightarrow P_i^{2s} = P_i^{-1}$  but just as doable. mod  $2 \rightarrow$  mod  $2s + 1$ .

## □ What about bosons?

Can be treated as infinitely-high-spin particles.

□ We thus have a method that automatically simulates essentially all condensed matter models.



*Particles of the standard model*

# PMR in quantum computing

# Same expansion but for quantum circuits

- In the context of quantum Hamiltonian simulation algorithms:

$$e^{-iHt}|z\rangle = \sum_{q=0}^{\infty} \sum_{\mathbf{i}_q} \underbrace{e^{-it[E_z, \dots, E_{z_q}]}_{\substack{\text{This is the divided-} \\ \text{difference exponential} \\ \text{whose inputs are the} \\ \text{classical energies along} \\ \text{the path.}}} \underbrace{D_{\mathbf{i}_q}}_{\substack{D_{\mathbf{i}_q} \text{ is the product} \\ \text{of all} \\ d_{i_j}(z_j) \text{ along path.} \\ \text{One for each} \\ \text{permutation} \\ \text{operator.}}} P_{\mathbf{i}_q}|z\rangle$$

Sum over all the combinations of products of  $P_j$  operators

$P_{\mathbf{i}_q} = P_{i_q} \dots P_{i_2} P_{i_1}$  (with  $q$  the size of the sequence)

- $\mathbf{i}_q = (i_1, i_2, \dots, i_q)$  is a multi-index. Each index  $i_j = 1 \dots M$  picks an off-diagonal permutation.

# Off-diagonal time evolution

- Consider from now on the short-time evolution operator

$$U = e^{-iH\Delta t}$$

- Using the off-diagonal series expansion we can write  $U$  as:

$$U = U \sum_z |z\rangle\langle z| = \sum_z U|z\rangle\langle z| = \sum_z V_z |z\rangle\langle z|$$

- Explicitly:

$$\begin{aligned} U &= \sum_z e^{-i\Delta t E_z} \sum_{q=0}^{\infty} \sum_{\mathbf{i}_q} e^{-i\Delta t [\Delta_z, \dots, \Delta_{z_q}]} D_{\mathbf{i}_q} P_{\mathbf{i}_q} |z\rangle\langle z| \\ &= \underbrace{\left( \sum_z \sum_{q=0}^{\infty} \sum_{\mathbf{i}_q} e^{-i\Delta t [\Delta_z, \dots, \Delta_{z_q}]} D_{\mathbf{i}_q} P_{\mathbf{i}_q} |z\rangle\langle z| \right)}_{\text{The off-diagonal evolution } U_{OD}} e^{-i\Delta t D_0} \end{aligned}$$

# The circuit

- Write the time-evolution operator as a product of short-time evolution operators.

$$e^{-iHt} = \underbrace{e^{-iH\Delta t} e^{-iH\Delta t} \dots e^{-iH\Delta t}}_{r \text{ times}}$$

- Time step is set to  $\Delta t = \ln 2 / \sum_{i=1} \Gamma_i$ .
- Important:  $r = t/\Delta t$ , number of repetitions is proportional to the “dimensionless time”  $T = t \sum_{i=1} \Gamma_i$  where  $\Gamma_i = |D_i|$  are the norms of the non-local off-diagonal operators.
- Diagonal part has been “integrated out”.



# Comparison with Taylor-based LCU

- Comparison with truncated Taylor series method is warranted.
- Gate cost of both algorithms scales linearly with their respective “dimensionless times”  $T$ . Qubit cost scales with their respective expansion orders  $Q \sim \log T$  and respective number of operators  $M$  to which the Hamiltonian is decomposed.

These numbers are very different.

Hamiltonian	$H = \sum_{ij} J_{ij} Z_i Z_j$	
Method	this paper	Taylor series LCU [1]
No. of LCU unitaries	0	$N^2$
Dimensionless time ( $T$ )	0	$t \sum_{ij}  J_{ij} $
Comments	$H$ is diagonal	-

Examples:

Hamiltonian	$H = \sum_{ij} J_{ij} Z_i Z_j + \sum_{ij} \tilde{J}_{ij} Z_i X_j$	
Method	this paper	Taylor series LCU [1]
No. of LCU unitaries	$N + 1$	$2N^2$
Dimensionless time ( $T$ )	$t \sum_j  \sum_i \tilde{J}_{ij} $	$t \sum_{ij} ( J_{ij}  +  \tilde{J}_{ij} )$
Comments	$D_0 = \sum_{ij} J_{ij} Z_i Z_j$ $D_j = \sum_i J_{ij} Z_i$	-

# Summary and conclusions

# Summary and conclusions

- We developed a very powerful classical simulation tool: **one algorithm to fit them all...**
- We have not solved the sign problem (but are working on it...).
- Haven't discussed measurements: We can show (paper in preparation) that essentially any conceivable physical operator may be measured, including integrated observables.
- **Method can be successfully ported to quantum computers.**

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IPAM workshop: Many-body Quantum Systems  
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