Sign problem and the Worldvolume Hybrid Monte Carlo method

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Based on work with

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Today's talk

- What is the sign problem?
 - Difficulty for complex action
 - Why the naive method (reweighting) fails
- <u>Overview of various approaches</u>
- Argue that <u>Worldvolume Hybrid Monte Carlo method</u> is a promising method

Introduction

Basics of Monte Carlo methods

What we want: expectation values of observables

 $\begin{cases} x = (x^{i}) \in \mathbb{R}^{N} : \text{dynamical variable } (N: \text{DOF}) \\ S(x): \text{ action} \\ \mathcal{O}(x): \text{ observable} \\ dx = \prod_{i=1}^{N} dx^{i} \\ \langle \mathcal{O} \rangle \equiv \int dx \ \rho(x) \mathcal{O}(x) \qquad \left(\rho(x) \equiv \frac{e^{-S(x)}}{\int dx e^{-S(x)}} \right) \end{cases}$

 $\begin{pmatrix} \text{e.g. scalar field} \\ x^i \leftrightarrow \phi(t, \mathbf{x}) \\ S(x) \leftrightarrow S[\phi] = \int dt \, d^3 \mathbf{x} \left[\frac{1}{2} (\partial_t \phi)^2 + \cdots \right] \\ dx = \prod_i dx^i \leftrightarrow [d\phi] = \prod_{t, \mathbf{x}} d\phi(t, \mathbf{x}) \end{pmatrix}$

When $N \gg 1$, numerical multiple integrals are unrealistic

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 $\rho(x)$

X

 \mathbb{R}^{N}

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When $N \gg 1$, numerical multiple integrals are unrealistic

However, when S(x) is real-valued, one can regard

$$\rho(x) = \frac{e^{-S(x)}}{\int dx \, e^{-S(x)}}$$

as a probability distribution $\begin{pmatrix} \because & \rho(x) \ge 0 & (\forall x) \\ \int dx \, \rho(x) = 1 \end{pmatrix}$

Markov chain Monte Carlo method

Basics of Monte Carlo methods

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Markov chain Monte Carlo method



What if the action is complex-valued?

<u>complex action</u> $S(x) = \operatorname{Re} S(x) + i \operatorname{Im} S(x) \in \mathbb{C} (\operatorname{Im} S(x) \neq 0)$

 $\rho(x)$ cannot be regarded as a probability distribution $\int \rho(x) = \frac{e^{-S(x)}}{\int dx e^{-S(x)}}$

■ <u>Examples</u>

$$\frac{\text{finite-density QCD}}{Z = \text{tr } e^{-\beta(H-\mu Q)}} \swarrow \qquad \begin{array}{l} Q: \text{ baryon number} \\ \mu: \text{ chemical potential} \\ = \int [dA_{\mu}][d\psi d\overline{\psi}] e^{(1/2g^2)\int \text{tr } F_{\mu\nu}^2 + \int \overline{\psi} D(A_{\mu};\mu)\psi} \\ = \int [dA_{\mu}][d\psi d\overline{\psi}] e^{(1/2g^2)\int \text{tr } F_{\mu\nu}^2} \int \text{Det } D(A_{\mu};\mu)\psi \\ = \int [dA_{\mu}] e^{(1/2g^2)\int \text{tr } F_{\mu\nu}^2} \int \text{Det } D(A_{\mu};\mu) \\ \in \mathbb{C} \quad \left(\because \left[\text{Det } D(A_{\mu};\mu) \right]^* = \text{Det } D(A_{\mu};-\mu^*) \right) \end{array}$$

- Quantum Monte Carlo (QMC)

strongly correlated electron systems, frustrated spin systems, ...

 $- \frac{\text{real-time dynamics of QM/QFT}}{\langle \Psi_2 | U(t_2, t_1) | \Psi_1 \rangle} = \int [dq(t)] \Psi_2^* (q(t_2)) \Psi_1 (q(t_1)) \underbrace{e^{iS[q(t)]}}_{\in \mathbb{C}} \underbrace{\left(U(t_2, t_1) = T \exp\left[-i \int_{t_1}^{t_2} dt H(t)\right] \right)}_{t}$



Plan

- 1. Introduction (done)
- 2. Sign problem and various approaches
- 3. Complex Langevin method
- 4. (Generalized) Lefschetz thimble method
- 5. Tempered Lefschetz thimble (TLT) method
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$$e^{-S(x)} = \underbrace{e^{-\operatorname{Re}S(x)}}_{\text{used for a wt}} \underbrace{e^{-i\operatorname{Im}S(x)}}_{\text{treated as part of obs}}$$

$$\langle \mathcal{O}(x) \rangle = \frac{\int dx \ e^{-S(x)} \mathcal{O}(x)}{\int dx \ e^{-S(x)}} = \frac{\int dx \ e^{-\operatorname{Re}S(x)} \ e^{-i\operatorname{Im}S_{I}(x)} \mathcal{O}(x) / \int dx \ e^{-\operatorname{Re}S(x)}}{\int dx \ e^{-\operatorname{Re}S(x)} \ e^{-i\operatorname{Im}S_{I}(x)} / \int dx \ e^{-\operatorname{Re}S(x)}}$$
highly oscillatory
$$= \frac{\langle e^{-i\operatorname{Im}S_{I}(x)} \mathcal{O}(x) \rangle_{\text{rewt}}}{\langle e^{-i\operatorname{Im}S_{I}(x)} \rangle_{\text{rewt}}}$$

$$= \frac{e^{-O(N)}}{e^{-O(N)}} \ (= O(1))$$

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$$= \frac{\langle e^{-i\operatorname{Im}S_{I}(x)} \, \mathcal{O}(x) \rangle_{\text{rewt}}}{\langle e^{-i\operatorname{Im}S_{I}(x)} \rangle_{\text{rewt}}} \left(\langle f(x) \rangle_{\text{rewt}} = \frac{\int dx \, e^{-\operatorname{Re}S(x)} \, f(x)}{\int dx \, e^{-\operatorname{Re}S(x)}} \right)$$

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■ **reweighting method**: simplest prescription for the complex action

$$e^{-S(x)} = \underbrace{e^{-\operatorname{Re}S(x)}}_{\text{used for a wt}} \underbrace{e^{-i\operatorname{Im}S(x)}}_{\text{treated as obs}}$$

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highly oscillatory = $\frac{\langle e^{-i\operatorname{Im}S_{I}(x)} \mathcal{O}(x) \rangle_{\text{rewt}}}{\langle e^{-i\operatorname{Im}S_{I}(x)} \rangle_{\text{rewt}}} \left(\langle f(x) \rangle_{\text{rewt}} \equiv \frac{\int dx \, e^{-\operatorname{Re}S(x)} f(x)}{\int dx \, e^{-\operatorname{Re}S(x)}} \right)$

$$= \frac{e^{-O(N)}}{e^{-O(N)}} (= O(1))$$

But, in numerical calc, the numer and denom are estimated with sample avgs

$$\langle \mathcal{O}(x) \rangle \approx \frac{e^{-O(N)} \pm O(1/\sqrt{N_{\text{conf}}})}{e^{-O(N)} \pm O(1/\sqrt{N_{\text{conf}}})} \qquad \begin{pmatrix} N : \text{DOF} \\ N_{\text{conf}} : \text{sample size} \end{pmatrix}$$

. small error $O(1/\sqrt{N_{\text{conf}}}) \leq e^{-O(N)} \Leftrightarrow \boxed{N_{\text{conf}} \geq e^{O(N)}}$
Need an exponentially large sample [4/36]

Example : Gaussian



Various approaches

■ <u>method 1</u> : no use of reweighting

- ▼ complex Langevin method [Parisi 1983, Klauder 1983] (may show a wrong convergence problem) (wrong results w/ small stat errors) method 2 : deformation of the integration surface Lefschetz thimble method [Witten 2010] [Cristoforetti et al. 2012, Fujii et al. 2013] generalized thimble method [Alexandru et al. 2015] tempered Lefschetz thimble method [MF-Umeda 2017, Alexandru et al. 2017] Worldvolume Hybrid Monte Carlo method [MF-Matsumoto 2020, Fujisawa et al. 2021, ...] ▼ path optimization method [Mori-Kashiwa-Ohnishi 2017, Alexandru et al. 2018] method 3 : no use of MC in the first place
 - ▼ tensor network (e.g. TRG) [Levin-Nave 2007, Shimizu-Kuramashi 2014,
 - good at calculating the free energy (but not so much for correl fcns)
 - direct treatement of Grassmann variables

Kadoh et al. 2020, ...]

Akiyama-san's talk

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Basic idea





Algorithm [Parisi 1983, Klauder 1983]

$$\begin{cases} \dot{z}_t = -\partial S(z_t) + v_t \\ z_{t=0} = x_0 \end{cases} \Rightarrow z_t = z_t(x_0; v) \\ v_t : \text{Gaussian white noise} \\ \left(\langle v_t v_{t'} \rangle_v = 2\delta(t - t') \right) \end{cases}$$

The following holds under a certain condition:

$$\int d^{2N} z \,\rho(z) \mathcal{O}(z) \stackrel{(*)}{=} \frac{\int_{\mathbb{R}^N} dx \, e^{-S(x)} \mathcal{O}(x)}{\int_{\mathbb{R}^N} dx \, e^{-S(x)}} \Big(= \langle \mathcal{O}(x) \rangle \Big)$$

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 $x \in \mathbb{R}^{N}$

Then, we have

$$\langle \mathcal{O}(x) \rangle \approx \frac{1}{N_{\text{conf}}} \sum_{k=1}^{N_{\text{conf}}} \mathcal{O}(z_{\tau+k})$$



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Wrong convergence problem

Condition for (*) to hold : [Aarts-James-Seiler-Stamatescu 1101.3270]

is not spread much in the direction $|\text{Im} z| \rightarrow \infty$

 $\rho_t(z \mid x_o): \begin{cases} \text{ lot optical index product index index index index index product index index$

Otherwise, it gives wrong estimates with small stat errors

Explicit form [Nagata-Nishimura-Shimasaki 1606.07627] Histogram of $|\partial S(x+iy)|$ decreases rapidly (at least exponentially)

Example : $e^{-S(x)} = (x + i\alpha)^4 e^{-x^2/2}$

[Nagata-Nishimura-Shimasaki 1606.07627]

 $\rho_t(z \,|\, x_0) \stackrel{t \to \infty}{\to} \rho(z)$

 X_0

 $z_t(x_0; v)$



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Basic idea of the thimble method (1/2)



(boundary at $|x| \rightarrow \infty$ kept fixed)

$$\langle \mathcal{O}(x) \rangle \equiv \frac{\int_{\Sigma_0} dx \, e^{-S(x)} \mathcal{O}(x)}{\int_{\Sigma_0} dx \, e^{-S(x)}} = \frac{\int_{\Sigma} dz \, e^{-S(z)} \mathcal{O}(z)}{\int_{\Sigma} dz \, e^{-S(z)}}$$

severe sign problem sign problem will be significantly reduced
if $\operatorname{Im} S(z)$ is almost constant on Σ

Basic idea of the thimble method (2/2)



Sign problem is expected to disappear on Σ_t at a sufficiently large t

Digression:thimble=YUBINUKI(指ぬき)



Digression:thimble=YUBINUKI(指ぬき)



How does the sign problem disappear?

• Integration on the original surface $\Sigma_0 = \mathbb{R}^N$ (flow time t = 0)



• Integration on a deformed surface Σ_t (flow time t)

$$\langle \mathcal{O}(x) \rangle = \frac{\langle e^{i\theta(z)} \mathcal{O}(z) \rangle_{\Sigma_t}}{\langle e^{i\theta(z)} \rangle_{\Sigma_t}} \approx \frac{e^{-e^{-\lambda t} O(N)} \pm O(1/\sqrt{N_{\text{conf}}})}{e^{-e^{-\lambda t} O(N)} \pm O(1/\sqrt{N_{\text{conf}}})}$$

$$\left(e^{i\theta(z)} \equiv e^{-i \operatorname{Im} S(z)} \frac{dz}{|dz|} \right)^{\Sigma_t} \left[e^{\lambda t} = O(N) \Leftrightarrow t = O(\log N) \right]$$

$$= \frac{O(1) \pm O(1/\sqrt{N_{\text{conf}}})}{O(1) \pm O(1/\sqrt{N_{\text{conf}}})}$$

$$\left[\lambda : \text{ (typical) singular value} \\ \text{of Hessian } \partial_i \partial_j S(\zeta) \right]$$

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Sign problem disappears at flow time $t = O(\log N)$

Example: Gaussian revisited

$$\begin{array}{l} \underline{Gradient flow}: \left[S(z) = (\beta/2)(z-i)^2 \Rightarrow S'(z) = \beta(z-i)\right] \\
\dot{z}_t = \overline{S'(z_t)} = \beta(\overline{z}+i) \text{ with } z_{t=0} = x \\
\Rightarrow z_t(x) = xe^{\beta t} + i(1-e^{-\beta t}) \quad \therefore |dz| = e^{\beta t} dx \\
\Rightarrow z_t(x) = xe^{\beta t} + i(1-e^{-\beta t}) \quad \therefore |dz| = e^{\beta t} dx \\
\Rightarrow \begin{cases} \operatorname{Re} S(z_t(x)) = \frac{1}{2}\beta e^{2\beta t} (x^2 - e^{-4\beta t}) \\ \operatorname{Im} S(z_t(x)) = -\beta x \quad \therefore e^{i\theta(z_t(x))} = e^{i\beta x} \end{cases} \quad \text{exponential growth} \\
\Rightarrow \langle f(z) \rangle_{\Sigma_t} = \frac{\int_{\Sigma_t} |dz| e^{-\operatorname{Re} S(z)} f(z)}{\int_{\Sigma_t} |dz| e^{-\operatorname{Re} S(z)}} = \frac{\int dx e^{-(\beta/2) e^{2\beta t} x^2} f(z_t(x))}{\int dx e^{-(\beta/2) e^{2\beta t} x^2}} \\
\end{array}$$

No small numbers appear

<

if we take a large t (=T) s.t. $e^{-\beta T} \ll \frac{1}{\sqrt{\beta}}$

$$\begin{aligned} x^{2} \rangle &= \frac{\langle e^{i\theta(z)} z^{2} \rangle_{\Sigma_{T}}}{\langle e^{i\theta(z)} \rangle_{\Sigma_{T}}} \\ &= \frac{e^{-(\beta/2)e^{-2\beta T}} \left(\beta^{-1} - 1\right)}{e^{-(\beta/2)e^{-2\beta T}}} = \frac{O(1)}{O(1)} \end{aligned}$$

<u>**NB</u></u>. Logarithmic increase is enough: T \sim O(\log \beta) (= O(\log N))</u>**



Appendix: Anti-holomorphic gradient flow

Behaviors around special points

(2) around a zero z_* of $e^{-S(z)}$: $e^{-S(z)} = (z - z_*)^{\gamma} f(z) \quad (\gamma \in \mathbb{Z}_{>0})$ $\Longrightarrow S(z) = -\gamma \ln(z - z_*) + \cdots$ $\Rightarrow \dot{z}_t = \overline{S'(z_t)} \approx -\gamma (\overline{z_t} - \overline{z_*})^{-1}$ $\Leftrightarrow \begin{cases} |z_t - z_*|^2 \approx |z_0 - z_*|^2 - 2\gamma t \\ \frac{z_t - z_*}{\overline{z_t} - \overline{z_*}} \approx \text{const} \end{cases}$ (converges to z_* in finite flow time)



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Ergodicity problem in thimble methods

large flow time t

relaxation of oscillatory integral

Sign problem resolved?

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Sign problem resolved? NO!

Actually, there comes out another problem at large *t* : **Ergodicity problem**



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Idea of tempering

[Marinari-Parisi 1992]

Suppose that the action $S(x;\beta)$ gives an ergodicity problem which disappears at a different value of β (say β_0)



Tempered Lefschetz thimble method [Fukuma-Umeda 1703.00861]

■ <u>TLT method</u>

- (1) Introduce replicas in between the initial integ surface $\Sigma_0 = \mathbb{R}^N$ and the target deformed surface Σ_T as $\left\{ \Sigma_{t_0=0}, \Sigma_{t_1}, \Sigma_{t_2}, \dots, \Sigma_{t_A=T} \right\}$
- (2) Setup a Markov chain for the extended config space $\{(t_a, x)\}$ (3) After equilibration, estimate observables with a subsample on Σ_T



Sign and ergodicity problems are solved simultaneously !

Hubbard model (1/4)

- Hubbard model toy model for electrons in a solid [Hubbard 1963]
- $c_{\mathbf{x},\sigma}^{\dagger}$, $c_{\mathbf{x},\sigma}$: creation/annihilation of an electron (site \mathbf{x} , spin $\sigma(=\uparrow,\downarrow)$)
- Hamiltonian

$$H = -\kappa \sum_{\langle \mathbf{x}, \mathbf{y} \rangle} \sum_{\sigma} c^{\dagger}_{\mathbf{x}, \sigma} c_{\mathbf{y}, \sigma} - \mu \sum_{\mathbf{x}} \left(n_{\mathbf{x}, \uparrow} + n_{\mathbf{x}, \downarrow} \right) + \frac{U}{2} \sum_{\mathbf{x}} n_{\mathbf{x}, \uparrow} n_{\mathbf{x}, \downarrow}$$

$$\begin{cases} n_{\mathbf{x},\sigma} \equiv c_{\mathbf{x},\sigma}^{\dagger} c_{\mathbf{x},\sigma} \\ \kappa(>0) : \text{hopping parameter} \\ \mu : \text{chemical potential} \\ U(>0) : \text{on-site repulsive potential} \end{cases}$$



• Quantum Monte Carlo (discretized imaginary time : $\beta = N_{\tau}\epsilon$)

Trotter decomposition + bosonization (HS transformation)

$$Z_{\beta,\mu} \equiv \operatorname{tr} e^{-\beta H}$$

$$\approx \int [d\phi] e^{-S[\phi_{\ell,\mathbf{x}}]} \equiv \int \prod_{\ell=1}^{N_{\tau}} \prod_{\mathbf{x}} d\phi_{\ell,\mathbf{x}} e^{-(1/2)\sum_{\ell,\mathbf{x}} \phi_{\ell,\mathbf{x}}^{-2}} \det M_{a}[\phi] \det M_{b}[\phi]$$

$$M_{a/b}[\phi] \equiv 1_{N_{s}} + e^{\pm\beta\mu} \prod_{\ell} \left(e^{\epsilon\kappa K} \operatorname{diag}[e^{\pm i\sqrt{\epsilon U}\phi_{\ell,\mathbf{x}}}] \right) : N_{s} \times N_{s} \operatorname{matrix}$$
[18/36]

Hubbard model (2/4)

[MF-Matsumoto-Umeda 1906.04243]



[19/36]

Hubbard model (3/4)



Hubbard model (4/4)

[MF-Matsumoto-Umeda 1906.04243]



When only a single (or very few) thimble is sampled by mistake, the average phase factor can take a larger value (due to the lack of cancellations among different thimbles)

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Pros and cons of the original TLT method

■ <u>TLT method</u> [MF-Umeda 2017]

Introduce replicas in between Σ_0 and Σ_T : $\left\{ \Sigma_{t_0=0}, \Sigma_{t_1}, \Sigma_{t_2}, \dots, \Sigma_{t_A=T} \right\}$



- <u>Pros</u>: can be applied to any systems once formulated by path integrals with continuous variables
- Cons : large comput cost at large DOF
 - necessary # of replicas $\propto O(N^{0-1})$
 - need to calculate Jacobian $J_t(x) = \partial z_t(x) / \partial x \propto O(N^3)$ everytime we exchange configs between adjacent replicas

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<u>Pros</u>: can be applied to any systems once formulated by path integrals with continuous variables

- \bigoplus major reduction of comput cost at large DOF
 - No need to introduce replicas explicitly
 - No need to calculate Jacobian $J_t(x) = \partial z_t(x) / \partial x$ in MD process
 - Configs can move largely due to the use of HMC

Worldvolume HMC (2/2) [MF-Matsumoto 2012.08468]

mechanism



Two pictures in WV-HMC

[MF-Matsumoto 2012.08468]

(1) <u>Target-space picture</u> [MF-Matsumoto 2012.08468]

sample: $\{z, z', z'', ...\}$

(2) <u>Parameter-space picture</u> [MF-Matsumoto 2012.08468] [Fujisawa et al. 2112.10519]

sample: $\{(t, x), (t', x'), (t'', x''), ...\}$



We employ (1) <u>target-space picture</u>



Computational cost of WV-HMC [MF-Matsumoto 2012.08468] $z = (z^i) \in \mathbb{C}^N$ ($N \propto V$: DOF) [MF-Matsumoto-Namekawa, Lattice2022] <u>Configuration flow</u> $\dot{z}_i = \overline{\partial_i S(z)} \implies O(N)$ <u>Vector flow</u> $\dot{\mathbf{v}}_i = \partial_i \partial_j S(z) \mathbf{v}_i \implies O(N^2)$ [when $\partial_i \partial_j S(z)$ is dense] $\frac{\Rightarrow O(N)}{\text{RATTLE}} \begin{bmatrix} z' = z + \Delta s \pi - \Delta s^2 \overline{\partial V(z)} - \lambda \end{bmatrix} \begin{bmatrix} \text{when } \partial_i \partial_j S(z) \text{ is sparse} \\ \text{(local field case)} \end{bmatrix}$ flow $\pi_{1/2} = (z' - z) / \Delta s$ $\tilde{\pi}' = \pi_{1/2} - \Delta s \,\overline{\partial V(z')}$ $\left(V(z) = \operatorname{\mathsf{Re}}S(z) + W(t(z))\right)$ \mathcal{U}_{\cdot} х $\pi' = \Pi'_{\mathcal{P}} \tilde{\pi}'$ \mathbb{C}^{N} cf) RATTLE on a single thimble $\mathcal{J} = \Sigma_{\infty}$ [Fujii et al. 2013] $z + \Delta z$ A. Σ_{T_1} RATTLE on Σ_t [Alexandru@Lattice2019, MF-Matsumoto-Umeda 2019] $\lambda \in N_z \mathcal{R}$ is determined s.t. $z' \in \mathcal{R}$ Σ_{t+h} For given $z = z_t(x)$ and π , find $h \in \mathbb{R}$, $u \in \mathbb{R}^N$, $\lambda \in N_z \mathcal{R}$ Σ_t \mathcal{R} flow s.t. $z_t(x) + \Delta s \pi - \Delta s^2 \overline{\partial V(z)} - \lambda = z_{t+h}(x+u)$ Σ_{T_0} x+uThis can be solved by Newton's method with BiCGStab for linear inversion (which requires only config/vector flows) $\Rightarrow O(N)$ Comput cost at each MD step is expected to be O(N)for local field theory (with the absence of fermion determinant)

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Appendix: Details of WV-HMC (1/2)



Appendix: Details of WV-HMC (2/2)

[MF-Matsumoto 2012.08468]

■ <u>Algorithm</u>

$$\mathcal{O}(x)\rangle = \frac{\langle A(z)\mathcal{O}(z)\rangle_{\mathcal{R}}}{\langle A(z)\rangle_{\mathcal{R}}} \left(\langle f(z)\rangle_{\mathcal{R}} \equiv \frac{\int_{\mathcal{R}} Dz \ e^{-V(z)}f(z)}{\int_{\mathcal{R}} Dz \ e^{-V(z)}}\right)$$

 $\begin{cases} V(z) = \operatorname{Re} S(z) + W(t(z)) &: \text{potential} \\ A(z) = \alpha^{-1}(z) e^{i\varphi(z)} e^{-i\operatorname{Im} S(z)} &: \text{reweighting factor} \end{cases} \begin{pmatrix} e^{i\varphi(z)} \equiv \frac{\det J}{|\det J|} \end{pmatrix}$

HMC on a constrained space [Andersen 1983, Leimkuhler-Skeel 1994]

 $\langle f(z) \rangle_{\mathcal{R}}$ is estimated with <u>RATTLE</u>

$$\begin{cases} \pi_{1/2} = \pi - \Delta s \,\overline{\partial V(z)} - \lambda \\ z' = z + \Delta s \,\pi_{1/2} \\ \pi' = \pi - \Delta s \,\overline{\partial V(z')} - \lambda' \end{cases}$$

 $\lambda \in N_z \mathcal{R}$ and $\lambda' \in N_{z'} \mathcal{R}$ are determined s.t.

 $\begin{cases} z' \in \mathcal{R} \\ \pi' \in T_{z'}\mathcal{R} \end{cases}$



cf) RATTLE on $\mathcal{J} = \Sigma_{\infty}$ [Fujii et al. 2013] RATTLE on Σ_t [Alexandru@Lattice2019, MF-Matsumoto-Umeda 2019]

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Successfully applied to ...

- (0+1)dim massive Thirring model [MF-Umeda 1703.00861] (TLT)
- 2dim Hubbard model [MF-Matsumoto-Umeda 1906.04243, 1912.13303] (TLT)
- chiral random matrix model (a toy model of finite-density QCD)
 [MF-Matsumoto 2012.08468] (WV-HMC)
- anti-ferro Ising on triangular lattice [MF-Matsumoto 2020, JPS meeting] (WV-HMC)
- complex scalar field at finite density [MF-Namekawa 2022, in preparation] (WV-HMC)

So far always successful for any models when applied, though the system sizes are not yet very large (DOF $N \lesssim 10^4$)

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Chiral random matrix model (1/2) [MF-Matsumoto 2012.08468]

■ <u>finite density QCD</u>

$$Z_{\text{QCD}} = \operatorname{tr} e^{-\beta(H-\mu N)} \begin{pmatrix} \{\gamma_{\mu}, \gamma_{\nu}\} = 2\delta_{\mu\nu}, \ \gamma_{\mu} = \gamma_{\mu}^{\dagger} = \begin{pmatrix} 0 & \sigma_{\mu} \\ \sigma_{\mu}^{\dagger} & 0 \end{pmatrix} \end{pmatrix}$$
$$= \int [dA_{\mu}] [d\psi d\overline{\psi}] e^{(1/2g^{2}) \int \operatorname{tr} F_{\mu\nu}^{2} + \int [\overline{\psi}(\gamma_{\mu}D_{\mu}+m)\psi + \mu\psi^{\dagger}\psi]}$$
$$= \int [dA_{\mu}] e^{(1/2g^{2}) \int \operatorname{tr} F_{\mu\nu}^{2}} \operatorname{Det} \begin{pmatrix} m & \sigma_{\mu}(\partial_{\mu} + A_{\mu}) + \mu \\ \sigma_{\mu}^{\dagger}(\partial_{\mu} + A_{\mu}) + \mu & m \end{pmatrix}$$
$$\text{toy model}$$

(

chiral random matrix model [Stephanov 1996, Halasz et al. 1998]

$$Z_{\text{Steph}} = \int d^2 W \ e^{-n \operatorname{tr} W^{\dagger} W} \det \begin{pmatrix} m & iW + \mu \\ iW^{\dagger} + \mu & m \end{pmatrix} \begin{pmatrix} \text{quantum field replaced by} \\ a \ \text{matrix incl spacetime DOF} \end{pmatrix} \\ (T = 0, N_f = 1) \end{pmatrix}$$

 $W = (W_{ij}) = (X_{ij} + iY_{ij}) : n \times n \text{ complex matrix}$ $\left(\mathsf{DOF} : N = 2n^2 \iff 4L^4(N_c^2 - 1)\right)$

role as an important benchmark model

- well approximates the qualitative behavior of QCD at large n
- complex Langevin suffers from wrong convergence [Bloch et al. 2018]





Finite-density complex scalar field (1/3)

$$\varphi(x) = \frac{1}{\sqrt{2}} [\varphi_1(x) + i\varphi_2(x)]$$
: complex scalar field

Continuum action

$$S(\varphi) = \int d^{d}x \Big[\partial_{\nu} \varphi^{*} \partial_{\nu} \varphi + m^{2} \varphi^{*} \varphi + \lambda (\varphi^{*} \varphi)^{2} + \mu (\varphi^{*} \partial_{0} \varphi - \partial_{0} \varphi^{*} \varphi) \Big]$$

$$\simeq \int d^{d}x \Big[(\partial_{\nu} \varphi^{*} + \mu \delta_{\nu,0} \varphi^{*}) (\partial_{\nu} \varphi - \mu \delta_{\nu,0} \varphi) + m^{2} |\varphi|^{2} + \lambda |\varphi|^{4} \Big]$$

Lattice action [Aarts 0810.2089]

$$S(\varphi) = \sum_{n} \left[(2d + m^2) |\varphi_n|^2 + \lambda |\varphi_n|^4 - \sum_{\nu=0}^{d-1} (e^{\mu \,\delta_{\nu,0}} \varphi_n^* \varphi_{n+\nu} + e^{-\mu \,\delta_{\nu,0}} \varphi_n \varphi_{n+\nu}^*) \right]$$

Introducing (ξ_n, η_n) with $\varphi_n = \frac{1}{\sqrt{2}}(\xi_n + i\eta_n)$, we have

$$S(\xi,\eta) = \sum_{n} \left[\frac{2d+m^{2}}{2} (\xi_{n}^{2}+\eta_{n}^{2}) + \frac{\lambda}{4} (\xi_{n}^{2}+\eta_{n}^{2})^{2} - \sum_{i=1}^{d-1} (\xi_{n+i}\xi_{n}+\eta_{n+i}\eta_{n}) - \cosh \mu (\xi_{n+0}\xi_{n}+\eta_{n+0}\eta_{n}) - i \sinh \mu (\xi_{n+0}\eta_{n}-\eta_{n+0}\xi_{n}) \right]$$

We complexify $(\xi, \eta) \in \mathbb{R}^{2V}$ to $(z, w) \in \mathbb{C}^{2V}$ with the flow equation $\dot{z}_n = [\partial S(z, w) / \partial z_n]^*, \quad \dot{w}_n = [\partial S(z, w) / \partial w_n]^* \begin{pmatrix} V : \text{ lattice volume} \\ \Rightarrow N = 2V \end{pmatrix}$

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Finite-density complex scalar field (2/3)

■ <u>Computational scaling in 2D</u>

[MF-Matsumoto-Namekawa, Lattice2022]

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The figure clearly shows that the comput cost scales as O(V) = O(N)

 $\begin{pmatrix} \mathsf{NB:} & \mathsf{The scaling will become } O(V^{1.25}) \\ & \mathsf{if we reduce the MD stepsize as } \Delta s \propto V^{-1/4} \\ & \mathsf{to keep the same amount of acceptance for increasing volume} \end{pmatrix}$

Finite-density complex scalar field (3/3)

[MF-Namekawa, in preparation]

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Comparison with TRG method [TRG (2D): Kadoh et al. 1912.13092] [TRG (4D): Akiyama et al. 2005.04645]



- WV-HMC gives results consistent with TRG at small volumes
- It should be interesting to investigte behaviors at large 4D volumes where TRG starts suffering from the systematic errors due to the introduction of $D_{\rm cut}$



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Summary and outlook

■ <u>Summary</u>

- ▼ WV-HMC seems to be a promising method esp. for local field theories
 - Allows a coding such as to enjoy the locality
 - Does not have a wrong convergence problem (as in the complex Langevin)
 - Can increase the precision to arbitrarily high order by increasing the sample size (No systematic errors such as those caused by introducing D_{cut} in TRG)

■ <u>Outlook</u>

- Application to QCD
 - WV-HMC for a path integration on a group manifold [MF, in preparation]
 - WV-HMC for pure Yang-Mills with finite θ [MF-Kanamori-Namekawa, ongoing]
 - WV-HMC for finite-density QCD [MF-Kanamori-Namekawa-..., ongoing]
- ▼ Further improvement of algorithm [MF, MF-Matsumoto-Namekawa-..., ongoing]
- ▼ Combining various algorithms

(e.g.) path optimization method and/or TRG (non-MC) (method and/or TRG (non-MC) (MF-Kadoh-Matsumoto 2107.14149)

▼ Particularly important: MC calc for time-dependent systems

first-principles calc of nonequilibrium processes such as those in the early universe, heavy ion collision experiments, ...

"Power of Monte Carlo"

Thank you.