Optimized flow for generalized Lefschetz thimble method

Katsuta Sakai (KEK \rightarrow Tokyo Medical and Dental University)

In collaboration with Jun Nishimura (KEK & SOKENDAI), Atis Yosprakob (Niigata University)

In preparation

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Monte-Carlo (MC) simulation..... An efficient method to have numerical integrations

Ex) path integral

$$\begin{split} \langle \mathcal{O} \rangle \equiv \frac{\int dx \, e^{-S(x)} \mathcal{O}(x)}{\int dx \, e^{-S(x)}} & \longrightarrow & \langle \mathcal{O} \rangle = \sum_{i=1}^{N_{\text{sample}}} \mathcal{O}(x^{(i)}) \\ & \text{With} \{x^{(i)}\}_i \text{ generated by } p(x^{(i)}) = \frac{e^{-S(x^{(i)})}}{\int dx e^{-S(x)}} \end{split}$$

weight \rightarrow probability dist.

MC with complex weight ($\, \, \Leftrightarrow S \in \mathbb{C}\,$)

$$\rightarrow \mathsf{Reweighting?} \quad \langle \mathcal{O} \rangle = \frac{\langle \mathcal{O}e^{-i\operatorname{Im}S} \rangle}{\langle e^{-i\operatorname{Im}S} \rangle}, \quad p(x^{(i)}) = \frac{e^{-\operatorname{Re}S(x^{(i)})}}{\int dx e^{-\operatorname{Re}S(x)}}$$

→ The phase $e^{-i \text{Im }S}$ oscillates; convergence is exponentially slow Sign problem : Obstacle to study on real-time dynamics, QCD with finite density and theta term, *etc*.

To circumvent the sign problem...

Generalized thimble method (GTM)

[Alexandru-Basar-Bedaque, '15] [Alexandru-Basar-Bedaque-Ridgway-Warrington, '16]

Complexify variables & deform integral contour with the holomorphic gradient eq.

Cf.) Picard-Lefschetz theory

$$x \to z(x,\sigma) \quad \text{with} \quad \frac{\partial z}{\partial \sigma} = \overline{\frac{\partial S(z)}{\partial z}}, \ \ z(x,0) = x$$

$$\frac{\partial S(z(x,\sigma))}{\partial \sigma} = \left|\frac{\partial S}{\partial z}\right|^2 > 0$$

Along the flow , $e^{-i{\rm Im}\,S}$: constant

$$e^{-\operatorname{Re}S}$$
: decrease

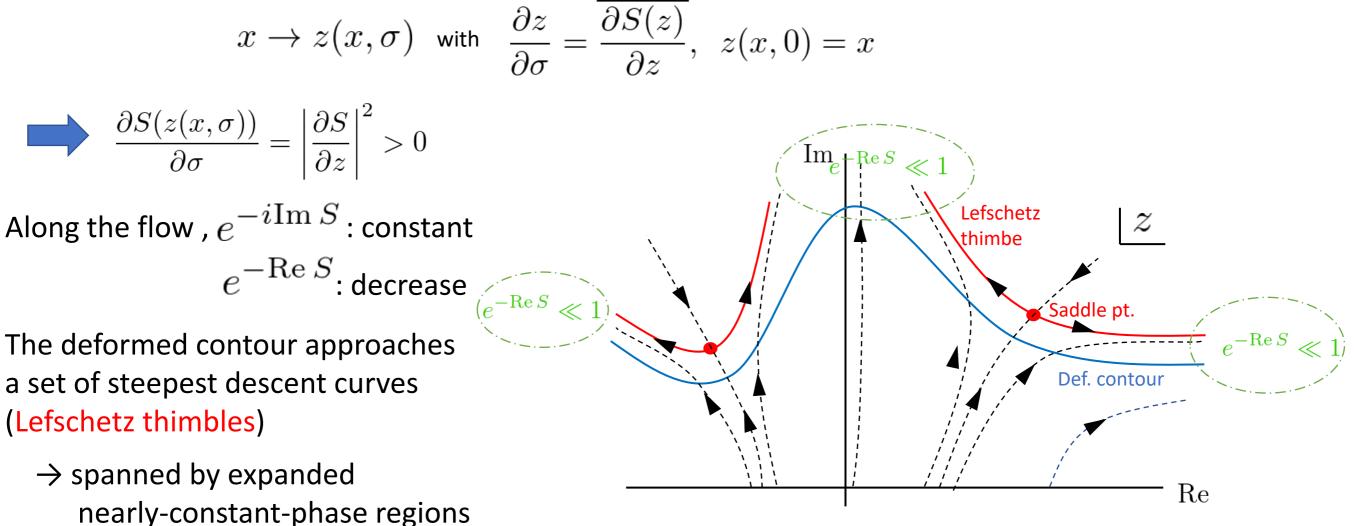
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Why expansive map?

• flow eq. for deviation $\ \delta z_j = z_j - z_j'$:

$$\frac{\partial (\delta z_j)}{\partial \sigma} = \overline{H}_{jk} \overline{\delta z_k}$$
 Hessian

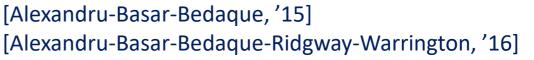
Roughly, exponential expansion with rates "singular values of \overline{H} "

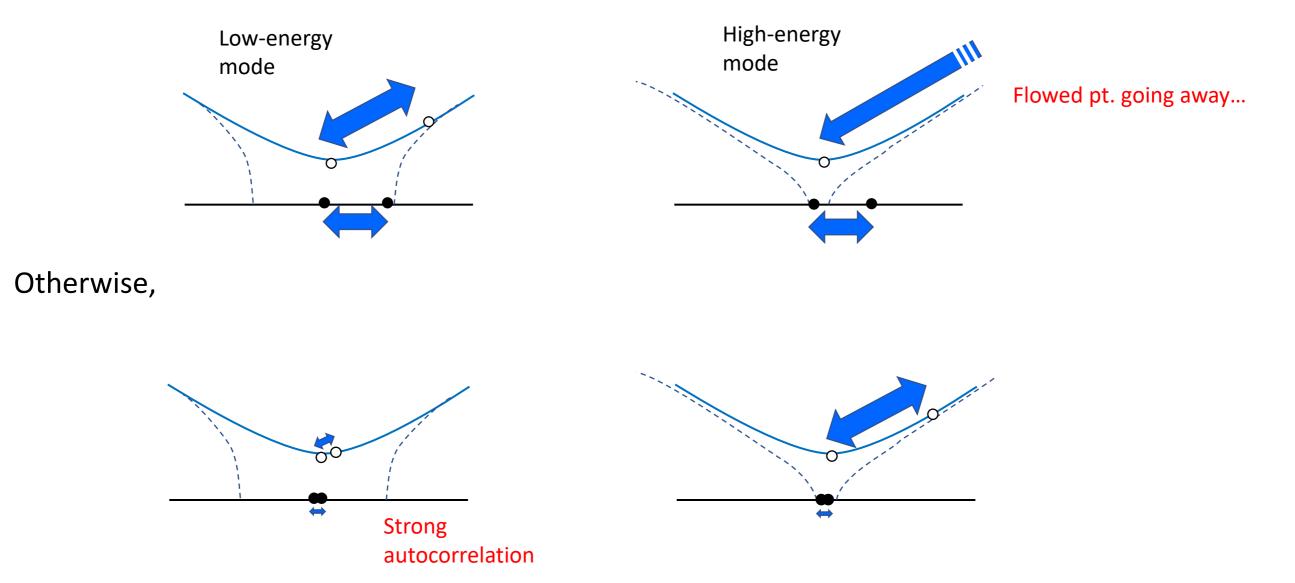
Subtlety in GTM

- Expansion rates differ mode by mode (high-energy mode → large expansion rate)
 Suitable region of flow time (to solving the sign prob. & Ergodicity prob.) also differ
- In the standard GTM, we move the sampling pt. on the original or deformed contour with mode-independent step size

Then, when we take the sampling with multimode...

Case a) Sampling point on the original contour

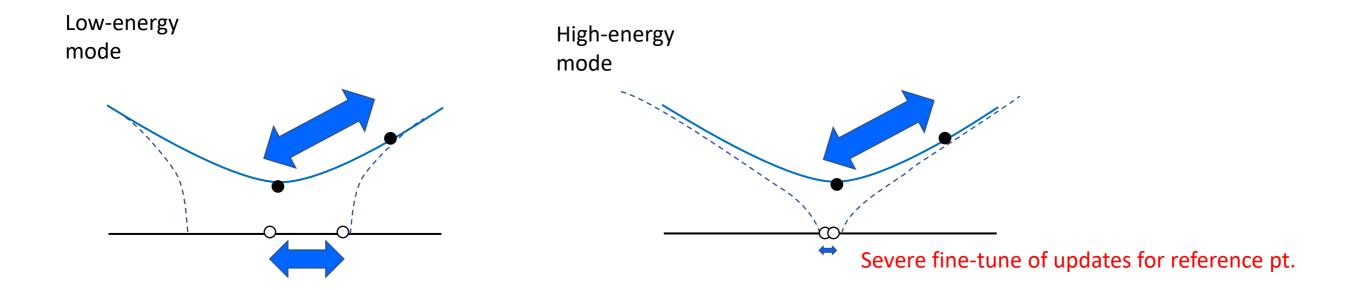




Then, when we take the sampling with multimode...

Case b) Sampling point on the deformed contour [Fukuma-Matsumoto, '20] Cf) [Fujii-Honda-Kato-Kikukawa-Komatsu-Sano, '13]

Note: a reference point and its update are required to compute the sampling pt.



2. Optimizing the flow

- Express the singular values of \overline{H} as $\ \lambda_1 < \lambda_2 < \cdots < \lambda_N$

The problematic hierarchy of the expansion modes is characterized by

 $\eta(\overline{H})\equiv rac{\lambda_N}{\lambda_1}$ Condition number (linear equation) or Stiffness ratio (differential equation)

- · In local theories, large $\,\eta\,$ appears mainly due to the derivative in the kinetic term
- Typically, $\eta(\bar{H}) = O(1000)$ even with $N = O(10) \rightarrow$ Averaging the flow time (at least) in Case a) is not so effective.
- In order to make the subtlety milder,

A) improve the numerical algorithm so that we can go with large $~\eta$

B) modify the flow eq. to make η close to 1 \leftarrow Our work

2. Optimizing the flow

Modified flow eq.

$$\frac{\partial z_j}{\partial \sigma} = A_{jk} \overline{\frac{\partial S}{\partial z_k}}, \quad A_{jk}(z, \bar{z}) : \text{hermitian, positive-definite}$$
"preconditioner"

Trade-off

• The essential property of the original flow eq. is inherited: $\frac{\partial S}{\partial \sigma} = \frac{\partial S}{\partial z_i} A_{jk} \frac{\partial S}{\partial z_k} > 0$

• Flow eq. for deviation : $\frac{\partial(\delta z_j)}{\partial \sigma} = A_{jk}\overline{H}_{kl}\overline{\delta z}_l + (\partial_l A_{jk})\overline{\partial_k S}\delta z_l + c.c.$ From interaction; small contribution to η in local theories From interaction; small contribution to η in local theories

We can freely choose $A_{jk}(z, \overline{z})$, but for the present purpose, it should be

- making $\eta(A\overline{H})$ as small as possible -
- · calculable at small cost

2. Optimizing the flow

One choice: full optimization as to $\,\eta\,$

$$A = (\overline{H}H)^{-\frac{1}{2}} \quad \Longrightarrow \quad \eta(A\overline{H}) = 1$$

Full preconditioner

Proof)

Polar decomposition: $\overline{H} = KU$ (K: Hermitian, positive-semidefinite U: unitary) $\Rightarrow \overline{H}H = H^{\dagger}H = K^{2}, \quad A = K^{-1}, \quad \eta(A\overline{H}) = \eta(U) = 1$

We have had a simulation with this choice for Case a), And investigated basic properties Samp

Sampling pt. on the original contour

Set up

Model: wavefunction of an anharmonic oscillator with Gaussian initial state

$$\psi_{\text{fin}}(b) = \int^{x(T)=b} \mathcal{D}x \, e^{-S} \psi_{\text{ini}}(x(0))$$
$$S = -i \int dt \left(\frac{\dot{x}^2}{2} - \frac{\lambda}{4!} x^4\right), \ \psi_{\text{in}}(x) = const. \times \exp\left(-\frac{\gamma}{4} (x-a)^2\right)$$

$$T = (N+1)\epsilon$$

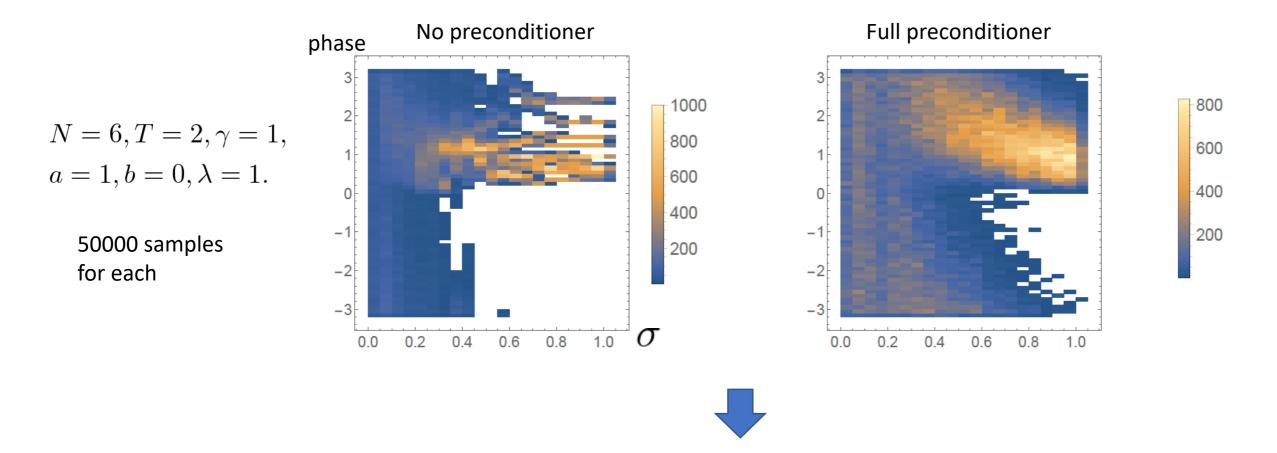
$$\psi_{\text{fin}}(b) = \int dx_1 \cdots dx_N e^{-S}$$

$$S = -i \sum_{i=0}^N \epsilon \left(\frac{(x_{i+1} - x_i)^2}{2\epsilon^2} - \frac{\lambda}{4!} \frac{x_{i+1}^4 + x_i^4}{2} \right) + \frac{\gamma}{4} (x_0 - a)^2, \quad x_{N+1} = b$$

• To move the sampling pt., we employ the backpropagating Hybrid Monte-Carlo [Fujisawa-Nishimura-KS-Yosprakob, '21]

HMC Update sampling pt. to distant place by "rolling down" the point with potential *S*. Backpropagation Fast calculation of the "force" to move the point for Case a); we solve the flow eq. backwardly.

Distribution of phase factor $e^{-i \operatorname{Im} S}$

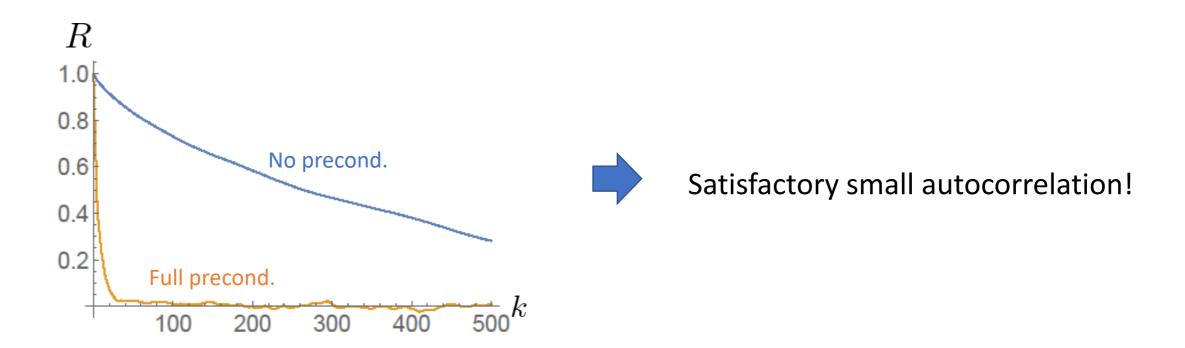


With the full preconditioner, we observed that the suitable regions of σ , where the sign prob. & Ergodicity prob. are resolved, are regulated among all the modes.

Autocorrelation

Between the samplings $x^{(i)}$ and $x^{(i+k)}$, the autocorrelation is defined as

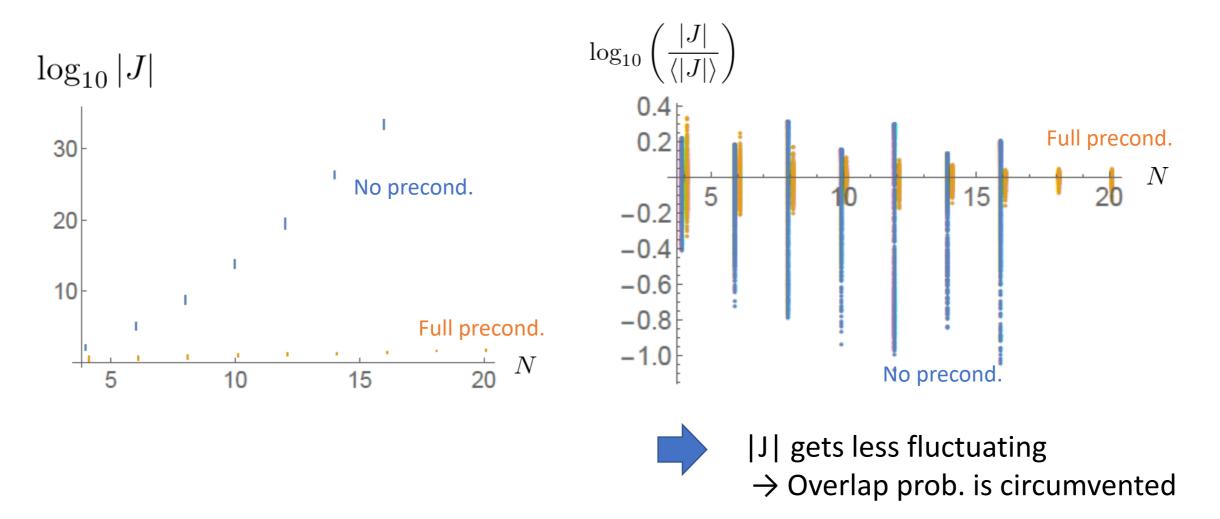
$$R(x;k) = \sum_{n=1}^{N-k} \frac{(x^n - \langle x \rangle) \cdot (x^{n+k} - \langle x \rangle)}{\langle x^2 \rangle - \langle x \rangle^2}$$



Fluctuation of Jacobian

In Case a), we have to include the Jacobian of the flow map in the observable.

 \rightarrow tend to fluctuate by orders of magnitude, causing an overlap problem.

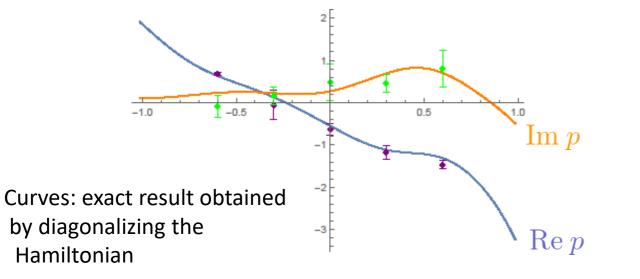


Larger system, stronger coupling (trial phase)

 $N = 20, \ \lambda = 30, \ \gamma = 4, \ a = 0.3$

Observable: momentum density

$$p(b) = \partial_b (\log \psi_{\text{fin}}(b)) = \frac{\partial_b \psi_{\text{fin}}(b)}{\psi_{\text{fin}}(b)} \qquad \left(\langle P \rangle = \int db \, |\psi_{\text{fin}}(b)|^2 \, p(b) \right)$$



Without preconditioner, it's very hard to obtain a result with a same-order precision in a realistic running time.

The applicability of GTM is enlarged, indeed!

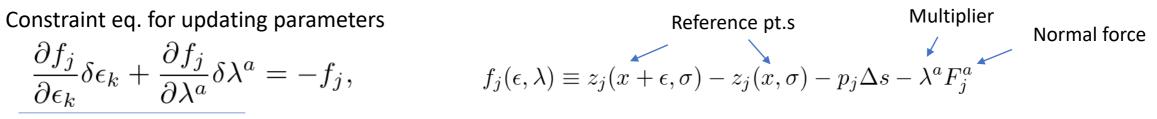
4. Summary and Future works

Summary

- We have discussed a subtlety in standard flow eq. in GTM, which comes from the hierarchy in the expansion rates. This is more crucial when we move the sampling pt. on the original contour.
- We have proposed a modification on the flow eq. by a preconditioner, which reduces the higherarchy.
- We have demonstrated a simulation for a real-time anharmonic oscillator, and have confirmed regulated suitable region of flowtime, small autocorrelation, small fluctuation of the Jacobian, enlarged applicability of GTM.

Future works

• Improvement on Case b), where we move the sampling pt. on the deformed contour?



Eq. with huge condition number \rightarrow reduced by a preconditioner

Or, world volume method [Fukuma-Matsumoto, '21, ...] (Prof. Fukuma's talk) with large-cond.num.-tolerant algorism?

- Other choices of the preconditioners, faster at the cost of $\eta \neq 1$?
- physical application such as quantum tunneling (Dr. Yosprakob's talk), quantum cosmology (Prof. Nishimura's talk)
 Listen to them tomorrow!

Back up

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Ex) Real time free particle $S = -i \int dt \frac{\dot{x}(t)^2}{2}$

$$\frac{\partial z(t;\sigma)}{\partial \sigma} = i \frac{\partial^2 \overline{z}(t;\sigma)}{\partial t^2}$$

Changing the variables: z = x + iy, u = x + y, v = x - y

$$\frac{\partial u}{\partial \sigma} = \frac{\partial^2 u}{\partial t^2}, \qquad \frac{\partial v}{\partial \sigma} = -\frac{\partial^2 v}{\partial t^2},$$
 Time-reversal diffusion eq.

Cf) Turok, "Existence of real time quantum path integrals" ('22)

