The path optimization for the sign problem of low dimensional QCD

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Sign problem

• When $S \in \mathbb{C}$, serious cancellation occurs in integration at large volume. And, integrals cannot be obtained precisely.

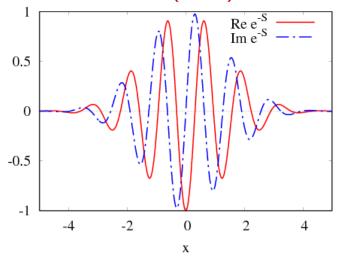
$$\mathcal{Z} = \int \mathcal{D}x e^{-\operatorname{Re}S - i\operatorname{Im}S} \ll \int \mathcal{D}x e^{-\operatorname{Re}S}$$

Seriousness of the sign problem
... average phase factor (APF)

$$APF = \frac{\int \mathcal{D}x e^{-\operatorname{Re}S} e^{-i\operatorname{Im}S}}{\int \mathcal{D}x e^{-\operatorname{Re}S}}$$
$$\sim e^{-\beta V\Delta f} \sim 0$$

ex.)
$$e^{-S(x)} = (x+10i)^{50}e^{-x^2/2}$$

J. Nishimura and S. Shimasaki, PRD 92 (2015) 011501



Path Optimization

<u>Y.M.</u>, K. Kashiwa, A. Ohnishi, Phys. Rev. D96 (2017) no.11, 111501

Optimize the integral path in the complexified variable space to weaken the sign problem.

(Integral of holomorphic(analytic) function is independent of integral path.)

We can regard the sign problem as an optimization problem.

(ex.) One variable case

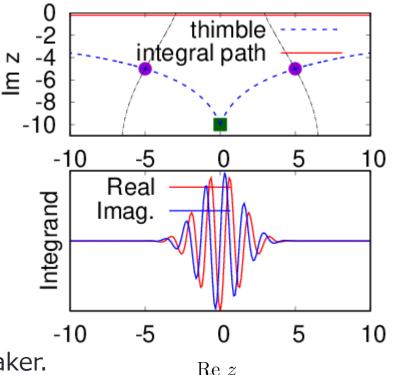
Trial function (integral path)

 $z(\cdot): \mathbb{R} \to \mathbb{C}$

Cost function (function to minimize)

 $\mathcal{F}[z(t)] = |\mathcal{Z}|\{|APF|^{-1} - 1\}$

oscillation of the integrand becomes weaker.



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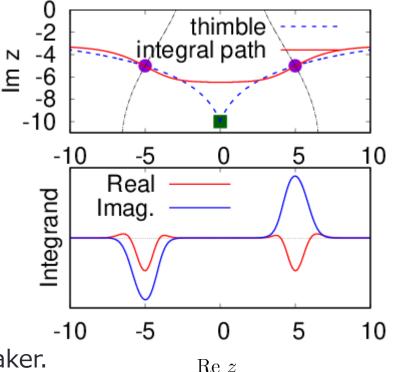
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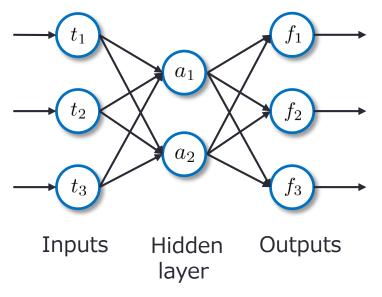
Neural Network for field theories

Neural Network(NN) is powerful to represent any functions of many inputs.

Combination of linear and non-linear transformation

 $a_{i} = g(W_{ij}^{(1)}t_{j} + b_{i}^{(1)})$ $f_{i} = g(W_{ij}^{(2)}a_{j} + b_{i}^{(2)})$ g(x) : Activation fn. (ex. tanh)

- $\bigotimes W, b, \alpha, \beta$: parameters
- Any fn. can be reproduced at (# of units of hidden layer) → ∞ (Universal approximation theorem) G. Cybenko, MCSS 2, 303 (1989) K. Hornik, Neural networks 4, 251 (1991)
- We input the real part of variables, and obtain the imaginary part from outputs.



 $z_i(t) = t_i + i(\alpha_i f_i(t) + \beta_i)$



YM, Kashiwa, Ohnishi, arXiv:1904.11140

Application to gauge theories

•1-spiecies of Staggered fermion LTM: Schmidt, et al.(2010) Di Renzo, et al.(2017)

$$S = \frac{1}{2} \sum_{\tau} (\overline{\chi}_{\tau} e^{\mu} U_{\tau} \chi_{\tau+\hat{0}} - \overline{\chi}_{\tau+\hat{0}} e^{-\mu} U_{\tau}^{-1} \chi_{\tau}) + m \sum_{\tau} \overline{\chi}_{\tau} \chi_{\tau}$$

$$\mathcal{Z} = \int \mathcal{D}U \det D[U] = \int dU \det [X_N + (-1)^N e^{\mu/T} U + e^{-\mu/T} U^{-1}]$$

 $X_N = 2\cosh(E/T), \ E = \operatorname{arcsinh} m, \ T = 1/N$

One link variable, No plaquette

0+1 dim. QCD

YM, Kashiwa, Ohnishi, arXiv:1904.11140

There are two ways of complexifying link variables.

① Complexification after diagonal gauge fixing

$$\begin{split} U &= \operatorname{diag}(e^{ix_1}, e^{ix_2}, e^{ix_3}), \ (x_3 &= -(x_1 + x_2)) \\ \mathcal{Z} &= \int dx_1 dx_2 \underline{H}(x) \exp(-S(x)) \\ & \operatorname{Haar\ measure} \\ x_a &\in \mathbb{R} \to z_a \in \mathbb{C} \\ \end{split} \begin{array}{l} & x_a (x) = x_a + iy_a(x), \ y_a(x): \text{ trial\ function} \\ \end{array} \end{split}$$

2 Complexification without diagonal gauge fixing U ∈ SU(3) → U ∈ SL(3, C)

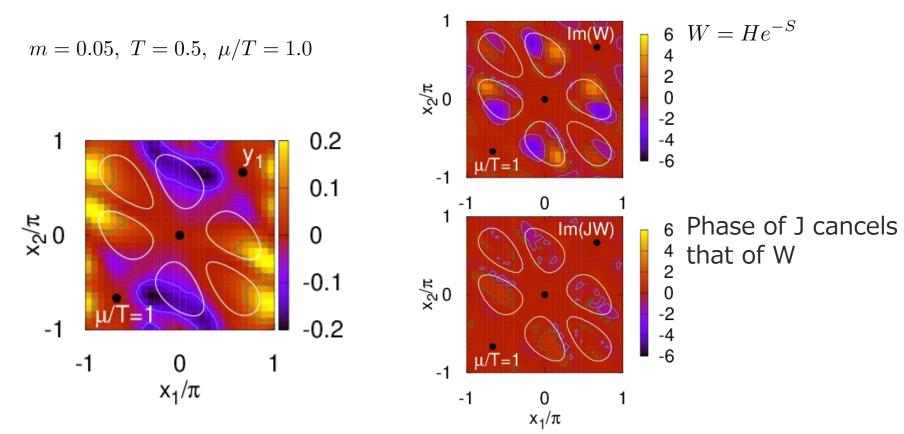
• ex.
$$\mathcal{U}(U) = U \prod_{a} \exp(y_a \lambda_a) \quad y_a(U)$$
 : trial function

1 dim. QCD diagonal gauge^{YM, Kashiwa, Ohnishi,} arXiv:1904.11140

① Complexification after diagonal gauge fixing

$$\mathcal{Z} = \int dz_1 dz_2 H(z) e^{-S(z)} = \int dx_1 dx_2 J H(z(x)) e^{-S(z(x))}$$

Optimization of z by gradient descent method



1 dim. QCD diagonal gauge^{YM, Kashiwa, Ohnishi,} arXiv:1904.11140

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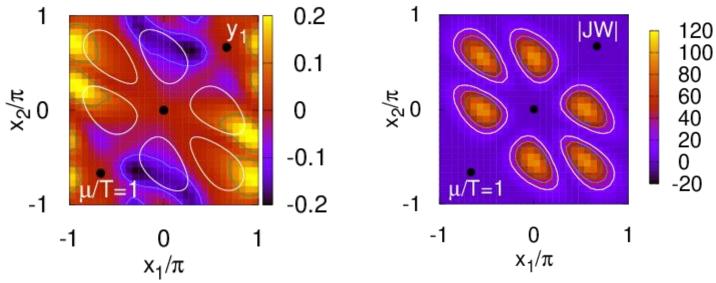
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Optimization of z by gradient descent method

$$m = 0.05, T = 0.5, \mu/T = 1.0$$

$$H(x) = \frac{8}{3\pi^2} \prod_{a < b} \sin^2\left(\frac{x_a - x_b}{2}\right)$$

→Separation of probability distribution. Difficulty in HMC.



1 dim. QCD one link

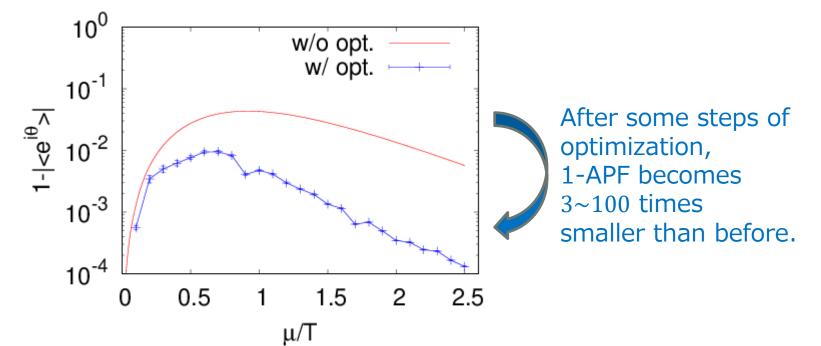
YM, Kashiwa, Ohnishi, arXiv:1904.11140

2 Complexification without diagonal gauge fixing

$$\mathcal{U}(U) = U \prod e^{y_a \lambda_a/2} \quad y_a(U)$$
 : trial function

We generate the configurations by HMC and optimize NN parameters by stochastic gradient descent(SGD, Adadelta).

•1 – average phase factor(APF)



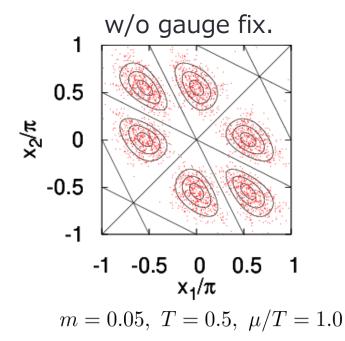
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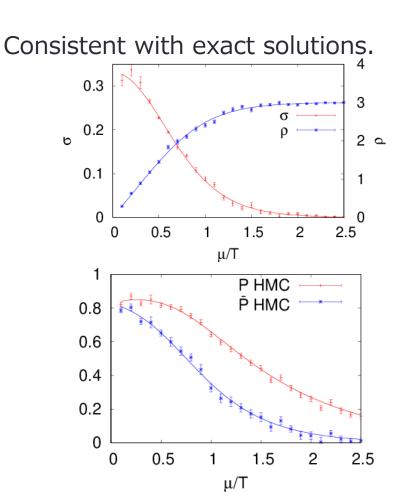
• Eigenvalue distribution

$$P\mathcal{U}P^{-1} = \text{diag}(e^{iz_1}, e^{iz_2}, e^{-i(z_1+z_2)})$$

Six separated region are produced by HMC



• Expectation value



1+1D QCD

 $S = \overline{\chi}_{x} D_{xy} \chi_{y}$

YM, Kashiwa, Ohnishi, in progress

- Action with staggered fermion (strong coupling)
 - $$\begin{split} D_{xy} &= m \delta_{x,y} + \frac{1}{2} (-1)^{x_0} \{ \delta_{x+\hat{1},y} U_{1,x} \delta_{x,y+\hat{1}} U_{1,y}^{-1} \} \\ &+ \frac{1}{2} \{ \delta_{x+\hat{0},y} e^{\mu} U_{0,x} \delta_{x,y+\hat{0}} e^{-\mu} U_{0,y}^{-1} \} \end{split}$$
- Polyakov gauge (without diagonal gauge fixing)

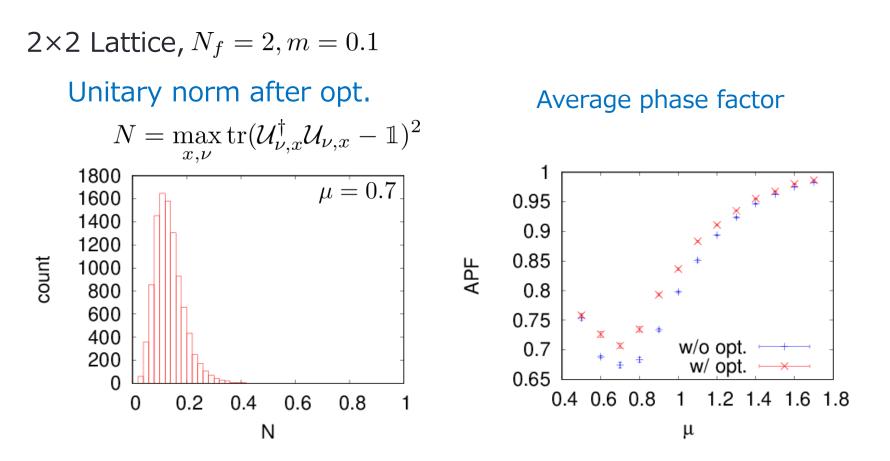
$$U_{0,x} \begin{cases} = 1 & (x_0 \neq N_\tau) \\ \in \mathrm{SU}(3, \mathbb{C}) & (x_0 = N_\tau) \end{cases}$$

Complexification

 $U_{\nu,x} \in SU(3) \to \mathcal{U}_{\nu,x}(U) \in SL(3,\mathbb{C})$



YM, Kashiwa, Ohnishi, in progress



Average phase factor can be slightly enhanced by the deformation of manifold. ...Is there upper bound of APF?

Summary

- In the path optimization method, we can regard the sign problem as an optimization problem.
 - Neural Network and optimization methods (SGD) developed in machine learning are helpful.
- We apply this method to 0+1 dim. QCD with and without diagonal gauge fixing.
 - Average phase factor becomes large and exact results are reproduced in observable calculations.
 - Without gauge fixing, eigenvalue distribution agrees with gauge fixed results, and HMC works well.
- In 1+1D QCD, works in progress.