

Fighting the sign problem with contour deformations

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22nd ZIMÁNYI SCHOOL WINTER WORKSHOP ON HEAVY ION PHYSICS
December 5-9, 2022, Budapest, Hungary

What is the sign problem?

Grand canonical partition function of QCD:

$$\begin{aligned} Z &= \int \mathcal{D}U \mathcal{D}\psi \mathcal{D}\bar{\psi} e^{-S[U, \psi, \bar{\psi}]} = \int \mathcal{D}U \det M[U] e^{-S_g[U]} \\ &= \int \mathcal{D}U e^{-S_{\text{eff}}[U]} = \int \mathcal{D}U w[U]. \end{aligned}$$

If $w[U]$ not real *and* positive $\leftrightarrow S_{\text{eff}} \notin \mathbb{R}$: MC with importance sampling not possible
 \sim **complex action problem.**

E.g.

- ▶ finite density/bariochemical potential QCD(-like models);
- ▶ Hubbard model of condensed matter physics;
- ▶ real time dynamics $\sim \langle f | e^{-iH} | i \rangle = \int \mathcal{D}x e^{iS[x]}$ (see previous talk by Paul Hotzy);
- ▶ etc.

What is the sign problem?

Q: How to overcome the complex action problem?

A: Simulate what you can *and* reweight to the original theory!

Expectation value through reweighting ($r[U]$ real and positive):

$$\langle \mathcal{O} \rangle_w = \frac{\int \mathcal{D}U \mathcal{O}[U] w[u]}{\int \mathcal{D}U w[u]} = \frac{\int \mathcal{D}U \mathcal{O}[U] \frac{w[U]}{r[U]} r[U]}{\int \mathcal{D}U \frac{w[U]}{r[U]} r[U]} = \frac{\langle \mathcal{O} \frac{w}{r} \rangle_r}{\langle \frac{w}{r} \rangle_r}.$$

Complex action problem reduces to the **sign problem**:

- ▶ large fluctuations in $\frac{w}{r}$ \rightarrow large cancellations \rightarrow large uncertainties (exp. in V, μ);
- ▶ severity of the sign problem:

$$\left\langle \frac{w}{r} \right\rangle_r = \frac{\mathcal{Z}_w}{\mathcal{Z}_r} \quad \Longrightarrow \quad \begin{cases} 1 & \sim \text{perfect!} \\ \approx 0 & \sim \text{not so much...} \end{cases}$$

E.g. *phase quenched* theory $r = |w| \sim \det M \rightarrow |\det M|$.

Fighting the sign problem?

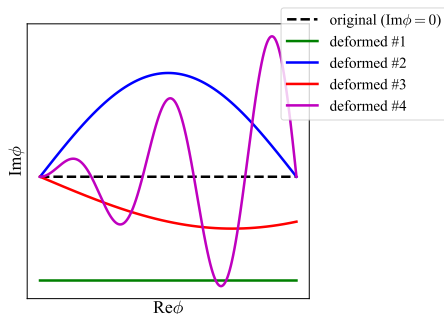
Why?:

- ▶ QCD;
- ▶ condensed matter physics (Hubbard model);
- ▶ neutron stars;
- ▶ hydrodynamic simulations at finite density;
- ▶ etc.

How?:

N -dim. integral over real fields to

- ▶ $2N$ -dim. integral over real and imaginary parts of complexified fields (e.g. *complex Langevin*);
- ▶ N -dim. integral with deformed integration contour/manifold into the complexified field space.



Complex contour deformations

Aim:

- ▶ searching for theories “closer” to the original theory;
- ▶ with real and positive weights;
- ▶ hence acquiring better signal-to-noise ratios in observables.

Set of integration manifolds $\mathcal{M}_{\text{def}}(\{p\})$ parameterised with some finite set of real parameters $\{p\}$:

$$\mathcal{Z} = \int_{\mathcal{M}_{\text{def}}} \mathcal{D}U_{\text{def}} w[U_{\text{def}}] = \int_{\mathcal{M}_{\text{def}}} \mathcal{D}X \det \mathcal{J}(X) w[U_{\text{def}}(X)].$$

In this case the phase quenched partition function:

$$\mathcal{Z}_{\text{PQ}}^{\text{def}}(\{p\}) = \int_{\mathcal{M}_{\text{def}}} \mathcal{D}X \left| \det \mathcal{J}(X) w[U_{\text{def}}(X)] \right|,$$

hence the severity of the sign problem:

$$\left\langle \frac{w}{r} \right\rangle_r = \frac{\mathcal{Z}}{\mathcal{Z}_{\text{PQ}}^{\text{def}}(\{p\})} = \left\langle \frac{\det \mathcal{J} w[U_{\text{def}}]}{|\det \mathcal{J} w[U_{\text{def}}]|} \right\rangle_{\text{PQ}}^{\text{def}} := \langle e^{i\theta} \rangle.$$

Integration manifold optimisation \sim machine learning

The sign problem is milder if

$$\frac{\mathcal{Z}}{\mathcal{Z}_{\text{PQ}}^{\text{def}}(\{p\})} \text{ is maximal!}$$

Y. Mori et. al. arXiv:1705.05605 [hep-lat]

Introducing a **cost function** and minimise it by varying $\{p\}$:

$$\mathcal{F}(\{p\}) = -\log\langle e^{i\theta} \rangle = -\log \mathcal{Z} + \log \mathcal{Z}_{\text{PQ}}^{\text{def}}(\{p\}) .$$

One can utilise machine learning algorithms (e.g. gradient descent) and compute gradients:

$$\nabla_p \mathcal{F}(\{p\}) = \nabla_p \log \mathcal{Z}_{\text{PQ}}(\{p\}) = -\langle \nabla_p S_{\text{eff}} - \nabla_p \log |\det \mathcal{J}| \rangle_{\text{PQ}}^{\text{def}} .$$

Application: Stephanov model

~ chiral random matrix model (Stephanov: [arXiv:hep-lat/9604003]):

$$\mathcal{Z} = \int \mathcal{D}U \det M[U] e^{-S_g[U]} \quad \text{vs.} \quad \mathcal{Z} = e^{N\mu^2} \int dW dW^\dagger \det^{N_f} (D + m) e^{-N \text{Tr}(WW^\dagger)},$$

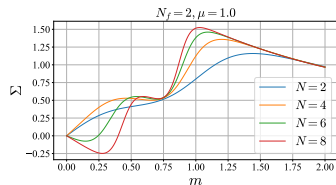
where:

- ▶ $W, W^\dagger \in \mathbb{C}^{N \times N}$, general complex matrices $\rightarrow 2N^2$ DoF;
- ▶ N_f : flavour number;
- ▶ μ : chemical potential;
- ▶ m : quark mass;
- ▶ and massless Dirac operator

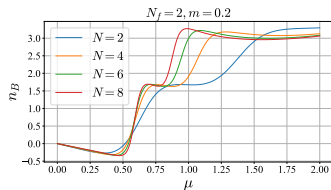
$$D = \begin{pmatrix} 0 & iW + \mu \\ iW^\dagger + \mu & 0 \end{pmatrix} \in \mathbb{C}^{2N \times 2N}$$

[arXiv:hep-ph/0003017]

- ▶ chiral condensate $\Sigma(m, \mu) \propto \partial \log \mathcal{Z} / \partial m$



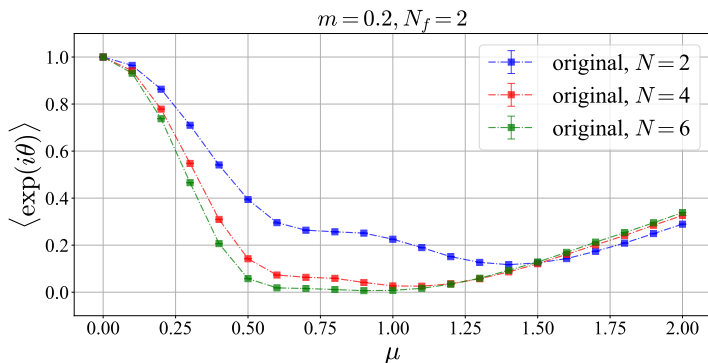
- ▶ baryon number density $n_B \propto \partial \log \mathcal{Z} / \partial \mu$



Application: Stephanov model and its sign problem at finite μ

Severity of the sign problem (average phase):

$$\langle e^{i\theta} \rangle = \frac{\mathcal{Z}}{\mathcal{Z}_{\text{PQ}}} = \left\langle \frac{\det^{N_f}(D+m)}{|\det^{N_f}(D+m)|} \right\rangle_{\text{PQ}}.$$



Complex Langevin method does not work for this model:

J. Bloch et. al arXiv:hep-lat/1712.07514.

Application: complexification and deformation *ansätze*

► Complexification:

$$W = A + iB \quad \rightarrow \quad X = \alpha + i\beta$$

$$W^\dagger = A^T - iB^T \quad \rightarrow \quad Y = \alpha^T - i\beta^T$$

$$A, B \in \mathbb{R}^{N \times N} \quad \text{and} \quad \alpha, \beta \in \mathbb{C}^{N \times N}.$$

α, β parameterised by A, B with some set of parameters $\{p\}$.

► Partition functions:

- deformations are chosen such that \mathcal{Z} remains invariant,
- while $\mathcal{Z}_{PQ} \equiv \mathcal{Z}_{PQ}(\{p\})$ does not!

► Motivation:

μ can be transformed out of the Dirac operator via a constant imaginary shift in matrix A :

$$D = \begin{pmatrix} 0 & (iA - B) + \mu \\ (iA^T + B^T) + \mu & 0 \end{pmatrix}$$



The *ansatz*:

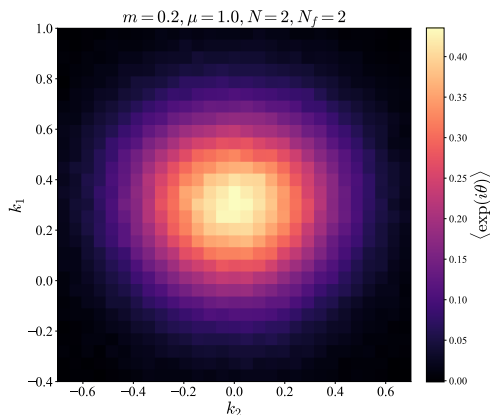
$$\alpha = A + ik_1 \text{id}$$

$$\beta = B + ik_2 \text{id}$$

$$k_1, k_2 \in \mathbb{R} \quad \text{and} \quad \det \mathcal{J} = 1.$$

Example result: constant shift *ansatz*

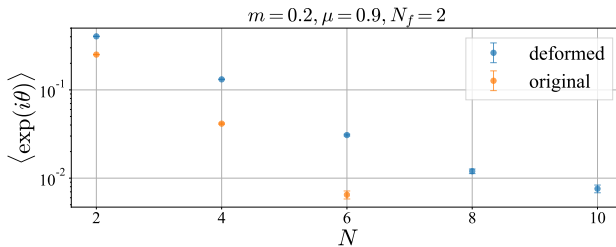
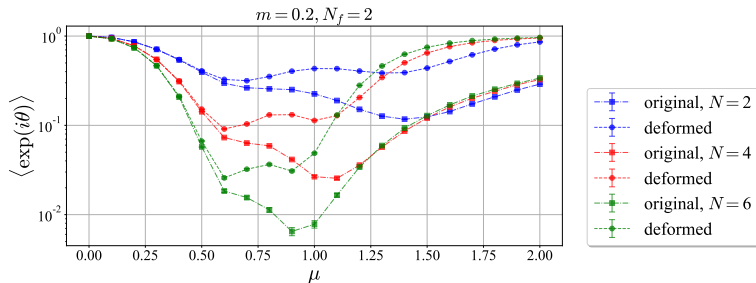
Ansatz:
 $\alpha = A + ik_1 \text{id}$
 $\beta = B + ik_2 \text{id}$



- ▶ Only relevant parameter is k_1 .
- ▶ Same result emerges from 20-parameter linear *ansatz* ($\text{Im}a = k_1$):

$$\alpha = (a + b\text{Tr}A + c\text{Tr}B)\text{id} + (1 + d)A + eB$$
$$\beta = (f + g\text{Tr}A + h\text{Tr}B)\text{id} + jA + (1 + k)B$$

Example result: μ - and N -dependence



Example result: piecewise optimisation of the trace

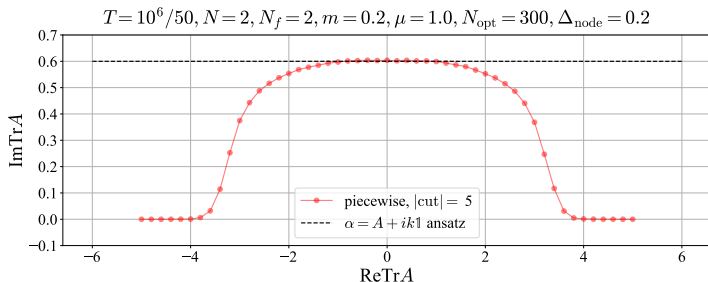
\sim deforming only $t = \text{Tr}A$.

Ansatz ($\beta = B$):

$$A = \frac{t}{N} \text{id} + \left(A - \frac{t}{N} \text{id} \right) = \frac{t}{N} \text{id} + \tilde{A} \quad \rightarrow \quad \alpha = \frac{\tau}{N} \text{id} + \tilde{A},$$

$\text{Tr}\tilde{A} = 0$ and $\tau = t + if(t; \{y_k\}, \{x_k\})$ where f is some (e.g. linear) interpolation function.

- ▶ $\{y_k\}$: parameters to optimise;
- ▶ $\{x_k\}$: nodes on the original contour.



Discussion and outlook

Findings:

- ▶ The sign problem in theories with a fermion determinant could be improved through complex contour deformations.
- ▶ Deformations that weaken the sign problem the most (i.e. some constant shift $\propto i \cdot \text{id}$) has no direct counterpart in full-QCD.
- ▶ Still, numerically the improvement appears to be exponential in V and μ .
- ▶ The optimisation method (i.e. machine learning) is an applicable way to find the optima of the deformation parameters in different ansätze.

To do:

- ▶ We shall use a more realistic toy model of QCD, or continue with chRMT but only with deformations allowed in full-QCD.
- ▶ Planned: applications in heavy dense QCD in 2 and/or 4 dimensions.

The End

Thank you for your attention.

Supported by the ÚNKP-22-3 New National Excellence Program of the Ministry for Culture and Innovation from the source of the National Research, Development and Innovation Fund