On the condition for correct convergence in the complex Langevin method

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arXiv:1606.07620 [hep-lat] and more

Complex Langevin Method (CLM)

[Parisi 83][Klauder 84]

Path integral for a theory with complex S

$$Z = \int_{-\infty}^{\infty} dx e^{-S(x)} \qquad \langle \mathcal{O}(x) \rangle = \frac{1}{Z} \int dx \mathcal{O}(x) e^{-S(x)}$$

· Langevin eq. for complexified variables $x \rightarrow z = x + iy \in \mathbf{C}$

$$\frac{dz(t)}{dt} = v(z) + \eta(t)$$

· drift term $v(z) = -\frac{dS(z)}{dz}$, gaussian noise η (t)

$$\lim_{t \to \infty} \langle \mathcal{O}(z_{\eta}(t)) \rangle_{\eta} = \langle \mathcal{O}(x) \rangle \qquad \langle \cdots \rangle_{\eta} = \int \mathcal{D}\eta \cdots e^{-\frac{1}{4} \int \eta(t')^2 dt'}$$

Key identity

[Aarts, Seiler, Stamatescu 09][Aarts, James, Seiler, Stamatescu 11]

$$\int dx dy \mathcal{O}(x+iy) P(x,y;t) = \int dx \mathcal{O}(x) \rho(x;t)$$
$$P(x,y;0) = \rho(x)\delta(y), \qquad \rho(x;0) = \rho(x)$$

• Probability distribution $P(x, y; t) = \langle \delta(x - x_{\eta}(t)) \delta(y - y_{\eta}(t)) \rangle_{\eta}$

•

 $\cdot \text{ complex weight } \rho \text{ (x;t) } \frac{\partial \rho(x;t)}{\partial t} = \frac{\partial}{\partial x} \left(\frac{dS}{dx} + \frac{\partial}{\partial x} \right) \rho(x;t) \quad \text{(Fokker-Planck eq)} \\ \rightarrow \text{e}^{-\text{S}} \text{ is a stationary solution}$

If the key identity holds and $\lim_{t\to\infty} P(x,y;t)$ converges to a unique function, then $\lim_{t\to\infty} \rho(x;t) = e^{-S}/Z$ and $\lim_{t\to\infty} \int dx dy \mathcal{O}(x+iy) P(x,y;t) = \frac{1}{Z} \int dx \mathcal{O}(x) e^{-S}$

Argument on the conditions for the CLM to work

 The previous argument for the justification of the CLM uses the continuous Langevin equation from the outset.

- Starting from a discretized Langevin equation, the $\varepsilon \rightarrow 0$ limit turns out to be subtle.
- The expectation value of the time-evolved observable $\mathcal{O}(z;t)$, which plays an important role in the argument, can be ill-defined.
- ☆ We fix these subtleties and establish the argument for the justification. This leads to a simple condition, which tells us whether the results are reliable or not.

Discretized Langevin equation

$$z^{(\eta)}(t+\epsilon) = z^{(\eta)}(t) + \epsilon z^{(\eta)}(t) + \sqrt{\epsilon}\eta(t)$$

- $\eta(t) = \eta^{(R)}(t) + i\eta^{(I)}(t)$ is a gaussian noise with probability $\langle \cdots \rangle_{\eta} \sim \int d\eta^{(R)}(t) d\eta^{(I)}(t) \cdots e^{-\frac{1}{4}\sum_{t} \left(\frac{1}{N_{R}}\eta^{(R)}(t)^{2} + \frac{1}{N_{I}}\eta^{(I)}(t)^{2}\right)}$ $N_{R} - N_{I} = 1$
 - Expectation value

$$\langle \mathcal{O}(z^{(\eta)}(t)) \rangle_{\eta} = \int dx dy \mathcal{O}(x+iy) P(x,y;t)$$

Infinitesimal (ε) time-evolution of the expectation value

$$\langle \mathcal{O}(z^{(\eta)}(t+\epsilon)) \rangle_{\eta} = \int dx dy \mathcal{O}_{\epsilon}(x+iy) P(x,y;t)$$

$$\mathcal{O}_{\epsilon}(z) = \int d\eta_R d\eta_I \, e^{-\frac{1}{4N_R}\eta_R^2 - \frac{1}{4N_I}\eta_I^2} \mathcal{O}\left(z + \epsilon \, v(z) + \sqrt{\epsilon} \left(\eta_R + i\eta_I\right)\right)$$

 ε -expansion and η -integration

$$\mathcal{O}_{\epsilon}(z) = \sum_{n=0}^{\infty} \frac{1}{n!} \epsilon^{n} : L^{n} : \mathcal{O}(z) \qquad L = \left\{ \operatorname{Rev}(z) + N_{R} \frac{\partial}{\partial x} \right\} \frac{\partial}{\partial x} + \left\{ \operatorname{Imv}(z) + N_{I} \frac{\partial}{\partial y} \right\} \frac{\partial}{\partial y} \\ : (f(x) + \partial_{x})^{2} := f(x)^{2} + 2f(x)\partial_{x} + \partial_{x}^{2}$$

• For holomorphic O(z)

$$L = \left\{ \operatorname{Re} v(z) + N_R \frac{\partial}{\partial x} \right\} \frac{\partial}{\partial x} + \left\{ \operatorname{Im} v(z) + N_I \frac{\partial}{\partial y} \right\} \frac{\partial}{\partial y}$$
$$\frac{\partial}{\partial x} \rightarrow \frac{\partial}{\partial z}, \quad \frac{\partial}{\partial y} \rightarrow i \frac{\partial}{\partial z} \qquad N_R - N_I = 1$$

$$\tilde{L} = \left(v(z) + \frac{\partial}{\partial z}\right)\frac{\partial}{\partial z}$$

 $LO(z)
ightarrow \tilde{L}O(z)$ holomorphic

$$\mathcal{O}_{\epsilon}(z) = \sum_{n=0}^{\infty} \frac{1}{n!} \epsilon^n : \tilde{L}^n : \mathcal{O}(z)$$

Subtlety

in the ε time-evolution

$$\langle \mathcal{O}(z^{(\eta)}(t+\epsilon)) \rangle_{\eta} = \int dx dy \mathcal{O}_{\epsilon}(x+iy) P(x,y;t) \qquad \tilde{L} = \left(v(z) + \frac{\partial}{\partial z} \right) \frac{\partial}{\partial z}$$
$$= \sum_{n=0}^{\infty} \frac{1}{n!} \epsilon^{n} \int dx dy \left\{ : \tilde{L}^{n} : \mathcal{O}(x+iy) \right\} P(x,y;t)$$

- The integrand involves the n-th power of the drift term
 → The prob. dist. of the drift should fall off faster than any power law at large magnitude
- If the integrals are convergent, we can take the $\varepsilon \rightarrow 0$ limit

$$\frac{d}{dt} \langle \mathcal{O}(z^{(\eta)}(t)) \rangle_{\eta} = \int dx dy \left\{ \tilde{L} \mathcal{O}(x+iy) \right\} P(x,y;t)$$

finite (τ) time-evolution and its subtlety Repeat the ε time-evolution $\tilde{L} = \left(v(z) + \frac{\partial}{\partial z}\right) \frac{\partial}{\partial z}$

$$\langle \mathcal{O}(z^{(\eta)}(t+\tau)) \rangle_{\eta} = \sum_{n=0}^{\infty} \frac{1}{n!} \tau^n \int dx dy \left\{ \tilde{L}^n \mathcal{O}(x+iy) \right\} P(x,y;t)$$

- · The infinite series should have a finite convergence radius
- The prob. dist. of the drift: $p(u;t) = \int dx dy \delta(u |v(z)|) P(x,y;t)$ • i.e. $p(u;t) \sim e^{-\kappa u}$ $\sum_{n=0}^{\infty} \frac{1}{n!} \tau^n \int_0^{\infty} du u^n p(u;t) = \sum_{n=0}^{\infty} \frac{1}{\kappa} \left(\frac{\tau}{\kappa}\right)^n$ convergence radius ~ κ

For the finite t-evolution of O(x) to be well-defined, the prob.
 dist. of the magnitude of the drift should fall off
 exponentially or faster

Proof of the key identity

 \cdot We prove the following identity for any k by induction wrt t

$$\int dxdy \left\{ \tilde{L}^k \mathcal{O}(x+iy) \right\} P(x,y;t) = \int dx \left\{ L_0^k \mathcal{O}(x) \right\} \rho(x;t)$$
$$\tilde{L} = \left(v(z) + \frac{\partial}{\partial z} \right) \frac{\partial}{\partial z} \qquad L_0 = \left(v(x) + \frac{\partial}{\partial x} \right) \frac{\partial}{\partial x}$$

· At t=0, it is trivially satisfied because $P(x,y;0) = \rho(x)\delta(y)$, $\rho(x;0) = \rho(x)$

· Assume this holds at some t, then for $\tau < \tau_{conv}$

$$\sum_{n=0}^{\infty} \frac{1}{n!} \tau^n \int dx dy \left\{ \tilde{L}^{n+k} \mathcal{O}(x+iy) \right\} P(x,y;t) = \sum_{n=0}^{\infty} \frac{1}{n!} \tau^n \int dx \left\{ (L_0)^{n+k} \mathcal{O}(x) \right\} \rho(x;t)$$

integration by parts
& FP eq for ρ (x;t)
$$\int dx dy \left\{ \tilde{L}^k \mathcal{O}(x+iy) \right\} P(x,y;t+\tau)$$
$$\int dx \left\{ (L_0)^k \mathcal{O}(x) \right\} \rho(x;t+\tau)$$

Demonstration of our condition in simple examples

· A model with a singular drift [Nishimura-SS 15]

$$Z = \int dx (x + i\alpha)^p e^{-\frac{1}{2}x^2} \qquad v(z) = \frac{p}{z + i\alpha} - z$$

A model with a possibility of excursions
 [Aarts-Giudice-Seiler 13]

$$Z = \int dx e^{-\frac{1}{2}(A+iB)x^2 - \frac{1}{4}x^4} \qquad v(z) = (A+iB)z - z^3$$

A model with a singular drift

[Nishimura-SS 15]

$$Z = \int dx (x + i\alpha)^p e^{-\frac{1}{2}x^2}$$

drift:
$$v(z) = \frac{p}{z + i\alpha} - z$$















A model with a possibility of excursions

[Aarts-Giudice-Seiler 13]

$$Z = \int dx e^{-\frac{1}{2}(A+iB)x^2 - \frac{1}{4}x^4}$$

drift:
$$v(z) = -(A + iB)z - z^3$$

potential danger of excursions in large |z|













Demonstration of our condition in Chiral Random Matrix Theory

CLM for Chiral Random Matrix Theory

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[Bloch-Bruckmann-Kieburg-Splitorff-Verbaarschot 13]

$$Z = \int d\Phi_1 d\Phi_2 d\Psi_1 d\Phi_2 [\det(D+m)]^{N_f} e^{-S} \qquad \text{[Mollgaard-Splittorff 13, 15]} \\ \text{[Nagata-Nishimura-SS 16]}$$

- For small masses, naive application of the CLM fails due to the singular drift problem (large drift) [Mollgaard-Splittorff 13]
 - By using a gauge cooling, we can avoid the problem and make the CLM work even at small masses [Nagata-Nishimura-SS 16]



Demonstration of our condition in CRMT $\Phi_1, \Phi_2, \Psi_1, \Psi_2 \rightarrow \text{drift } F_i \quad (i=1,2,3,4)$ eigenvalues of $\sqrt{F_i^{\dagger}F_i} \quad (i=1,2,3,4)$

without cooling (blue points)



Demonstration of our condition in CRMT $\Phi_1, \Phi_2, \Psi_1, \Psi_2 \rightarrow \text{drift } F_i \quad (i=1,2,3,4)$ eigenvalues of $\sqrt{F_i^{\dagger}F_i} \quad (i=1,2,3,4)$

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with a cooling (red points)



Summary

· We have established the argument for justification of the CLM by starting with a finite step-size ϵ

 We find that in order for the CLM to work, the probability distribution of the drift should fall off exponentially or faster.

 We demonstrate the validity of our condition in simple models involving the singular drift problem or the excursion problem and Chiral Random Matrix Theory.

Importantly, our condition turns out to be practically useful in the CLM for lattice QCD \rightarrow Nagata's talk (next)