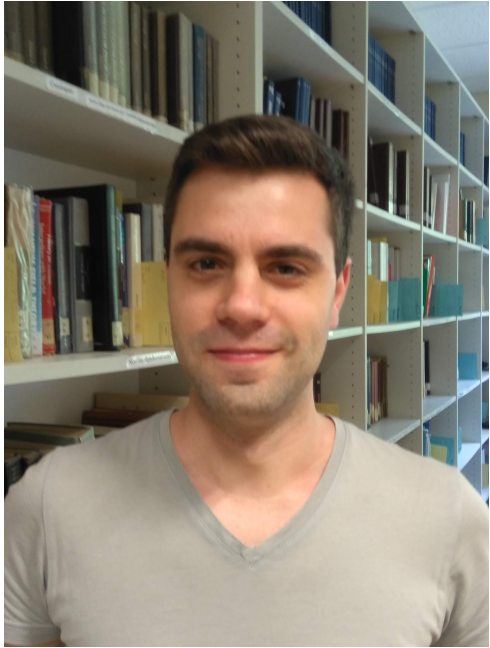


**ALLEVIATING THE SIGN PROBLEM WITH HOLOMORPHIC FLOW:
AN EXERCISE IN MACHINE LEARNING**
TBILISI, GEORGIA, 09.2019

THOMAS LUU, IKP-3/IAS-4, FZJ/ HSKP, UNIVERSITÄT BONN

MY PARTNERS IN CRIME . . .



Jan-Lukas Wynen



Evan Berkowitz



Christopher Körber

Shout outs also go to:

Stefan Krieg, Timo Lähde, Johann Ostmeyer, Marcel Rodekamp, Carsten Urbach, Xiaonu Xiong

IF ONLY THERE WASN'T ANY SIGN PROBLEM...

- Path-integral formalism

$$\begin{aligned}\langle \hat{O} \rangle &= \frac{1}{Z} \int_{\phi \in \mathbb{R}^N} \mathcal{D}[\phi] e^{-S[\phi]} O[\phi] \\ &= \int_{\phi \in \mathbb{R}^N} \mathcal{D}[\phi] \mathbb{P}[\phi] O[\phi]\end{aligned}$$

“Probability density”

Note: If $S[\phi] = S_R[\phi] + i S_I[\phi]$ then $\mathbb{P}[\phi] \mathcal{D}[\phi] \notin [0, \infty)$

Question: How do we interpret such “Probabilities”?

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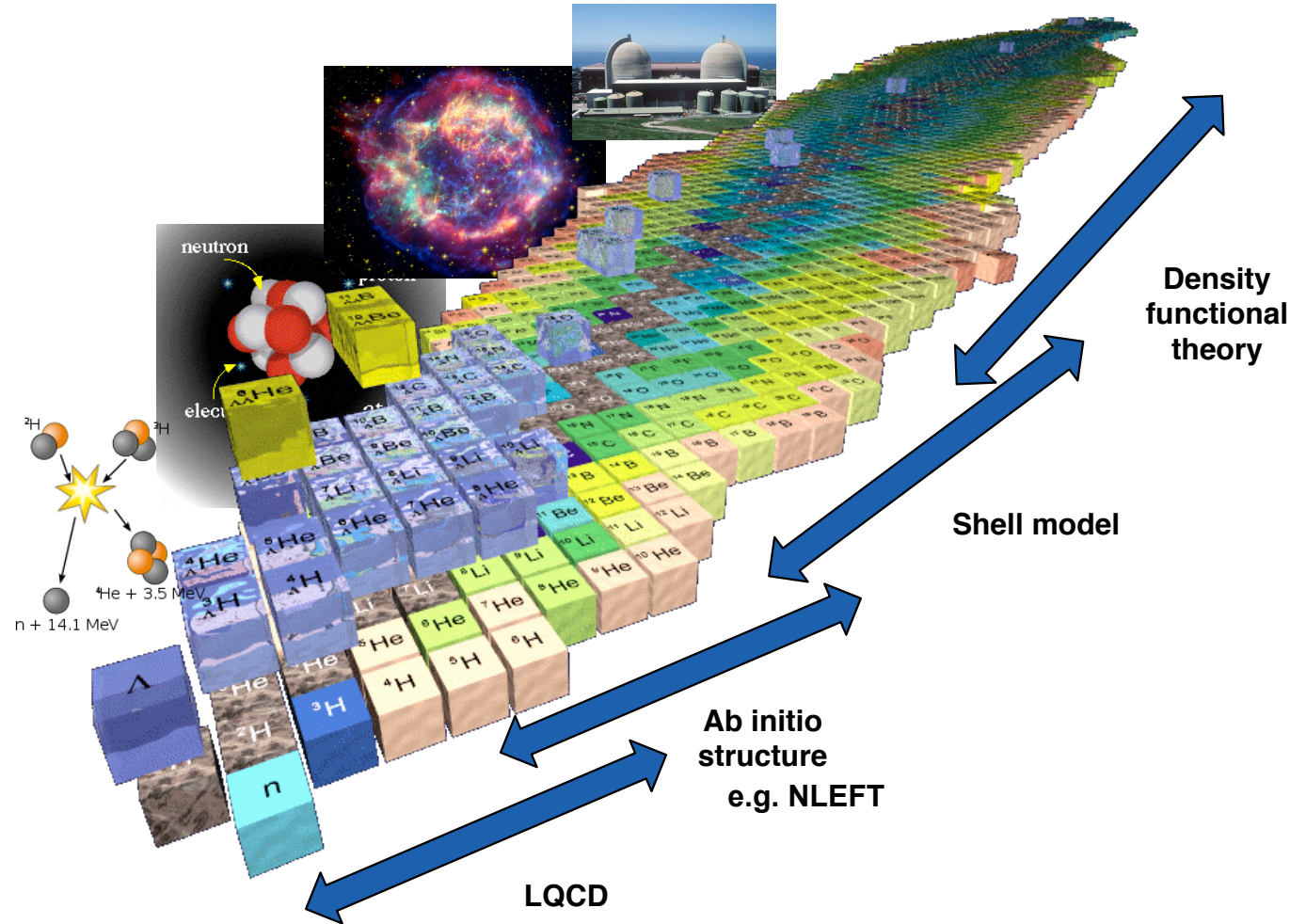
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“Complex Phase sign problem”

WHEN DOES THE SIGN PROBLEM SHOW UP?

- Life in Minkowski space
- Finite density field theory
 - What are the phases of quark matter at high density?
 - Baryon chemical potential
- Complex term in Lagrangian (even after Wick rotation)
 - QCD theta term
- Stochastic calculations of neutron rich nuclei

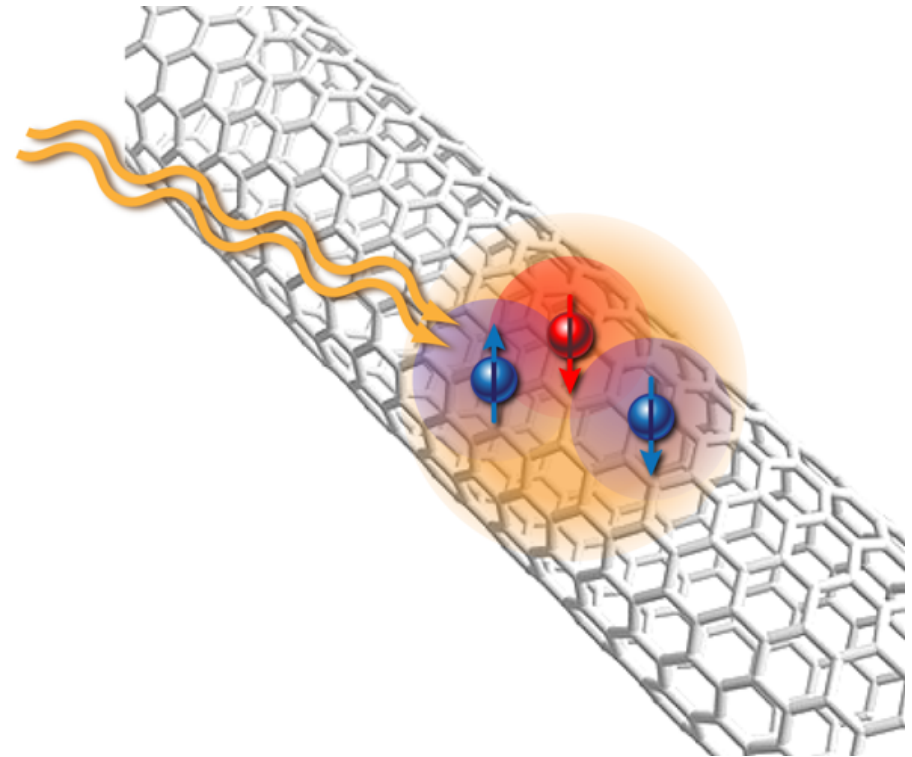


Sign problem shows up in the *most interesting problems in nuclear physics!!*

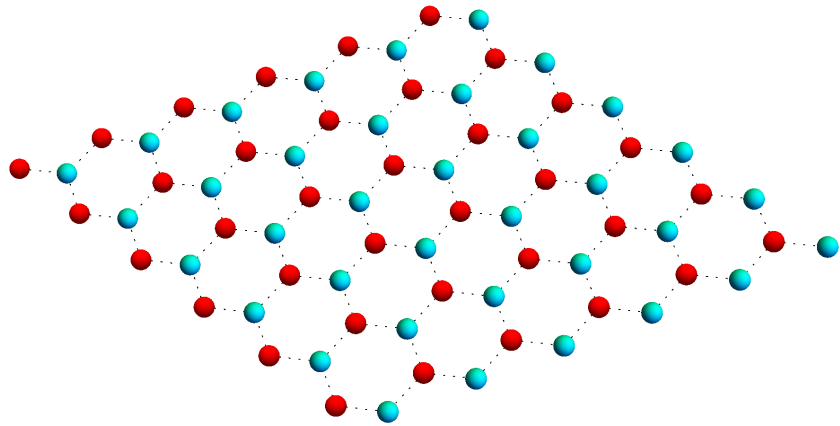
BUT THE SIGN PROBLEM IS NOT UNIQUE TO NUCLEAR PHYSICS...

Condensed matter/solid state communities also share our pain...

- High density strongly correlated electrons
- Doped systems
 - Electron chemical potential

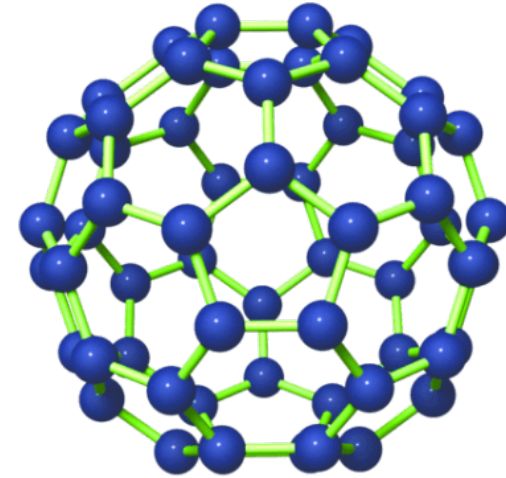


THE TYPE OF THE LATTICE ALSO CAN INFLUENCE THE SIGN PROBLEM



bipartite lattice

$$\mathbb{P}[\phi] \propto \det (M[\phi]M^\dagger[\phi]) \geq 0$$

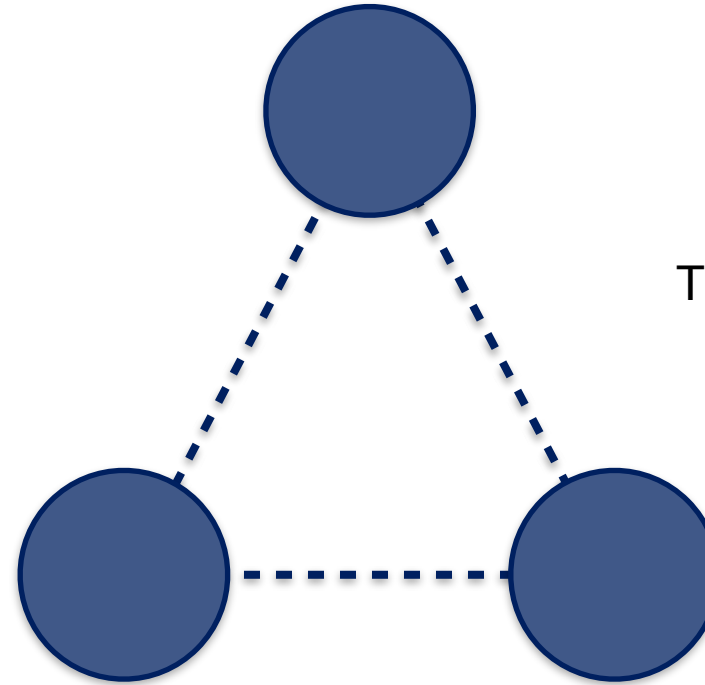


non-bipartite lattice

$$\begin{aligned} \mathbb{P}[\phi] &\propto \det (M[\phi, \kappa]M[-\phi, -\kappa]) \\ &\in (-\infty, \infty) \end{aligned}$$

THE (NEXT TO) SIMPLEST NON-BIPARTITE LATTICE: THE TRIANGLE

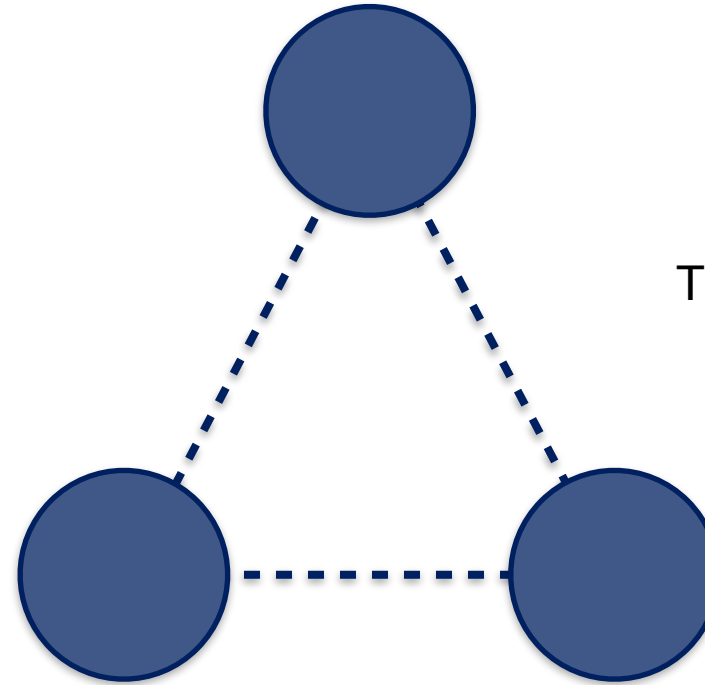
- Can be solved exactly via direct diagonalization
 - Can immediately test new methods
- Severity of sign problem controlled by strength of interaction and temperature of system
- No need to introduce chemical potential



Three-site problem

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Numerical examples shown in this talk will be done on this system

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 - Discuss in the next slide

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 - Expand in μ (baryon chemical potential)
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 - When it’s wrong, it’s really wrong

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Holomorphic flow (Lefschetz Thimbles)
The subject of this talk. . .

HUBBARD MODEL ON A NON-BIPARTITE LATTICE

Introducing the Hamiltonian

$$H = - \sum_{x,y} (a_x^\dagger \kappa_{xy} a_y + b_x^\dagger \kappa_{xy} b_y) + \frac{U}{2} \sum_x (n_x - \tilde{n}_x)^2$$

$$Z = \text{Tr } e^{-\beta H}$$

Goal: Apply PI
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Integrate out Grassmann variables

... and so on ...

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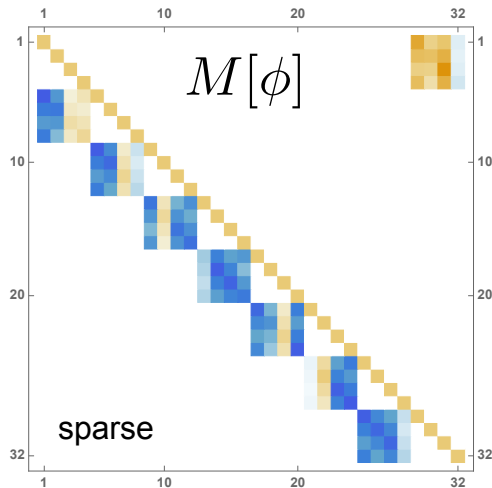
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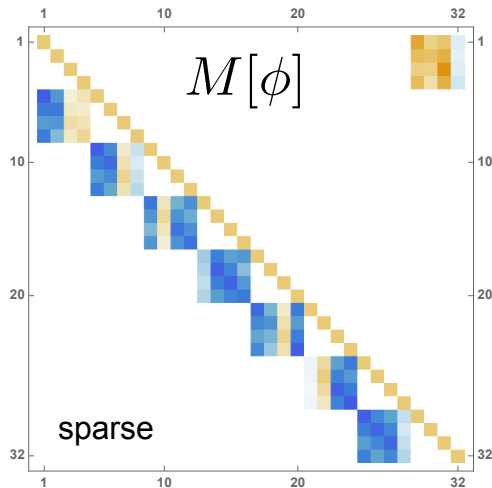
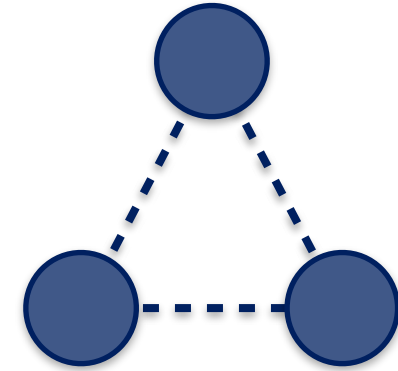
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RE-WEIGHTING IS THE STANDARD “GO-TO” METHOD

$$\mathbb{P}[\phi] = \det (M[\phi, \kappa] M[-\phi, -\kappa]) e^{\left(-\frac{1}{2\tilde{U}} \sum_{x,t} \phi_{xt}^2\right)} = e^{-S_R[\phi]} e^{-iS_I[\phi]}$$

No longer positive definite! = “sign problem”

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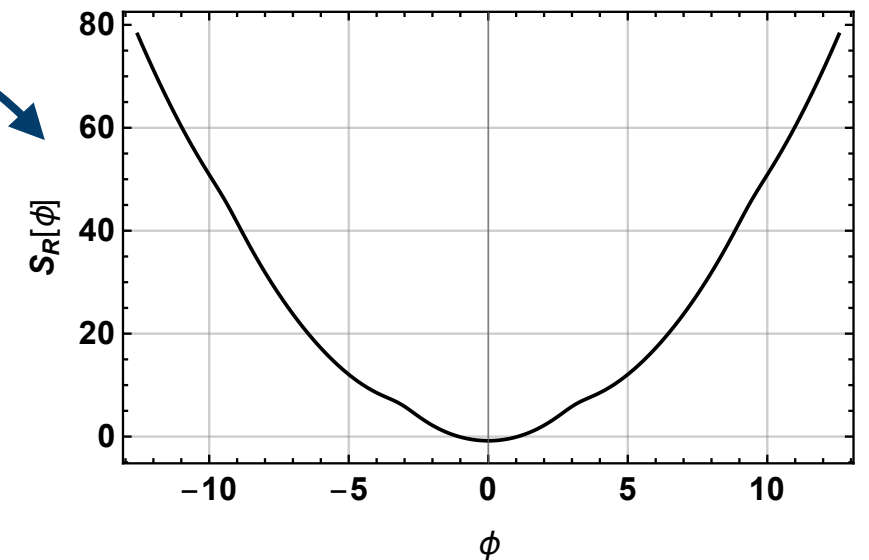
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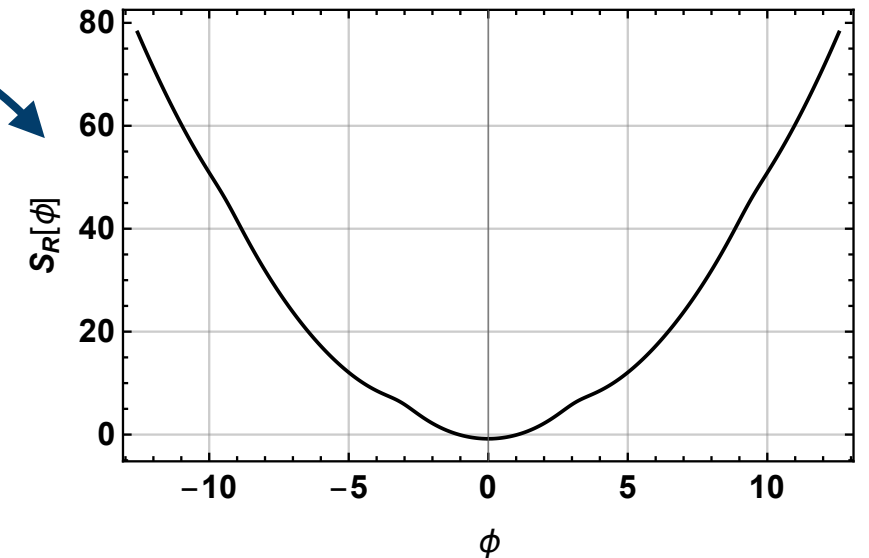
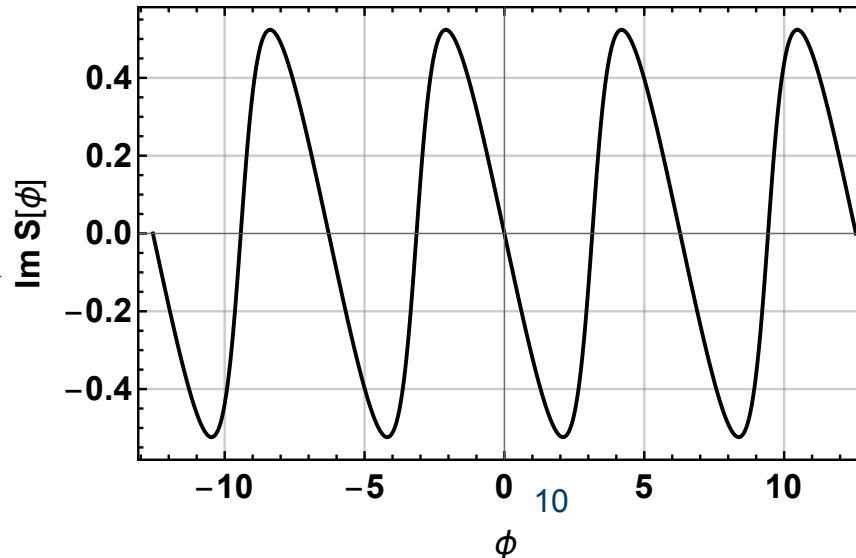
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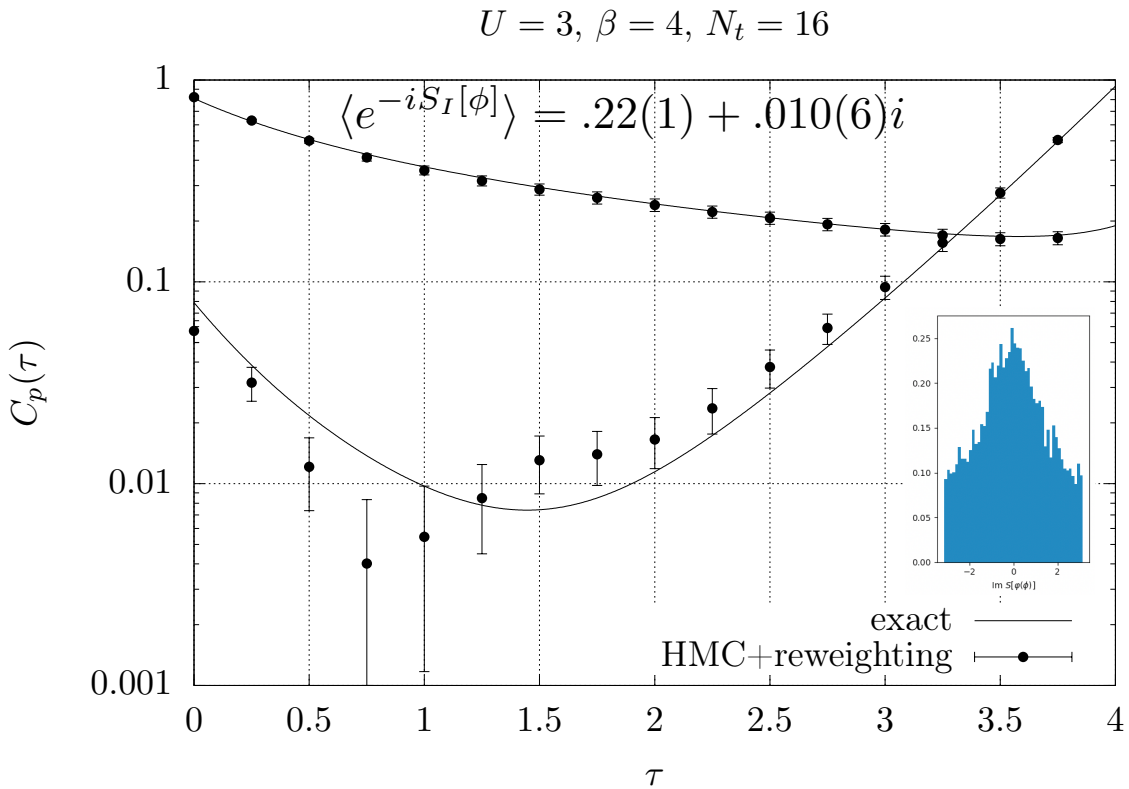
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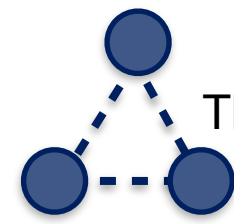
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RE-WEIGHTING WORKS GREAT WHEN IT WORKS, BUT. . .

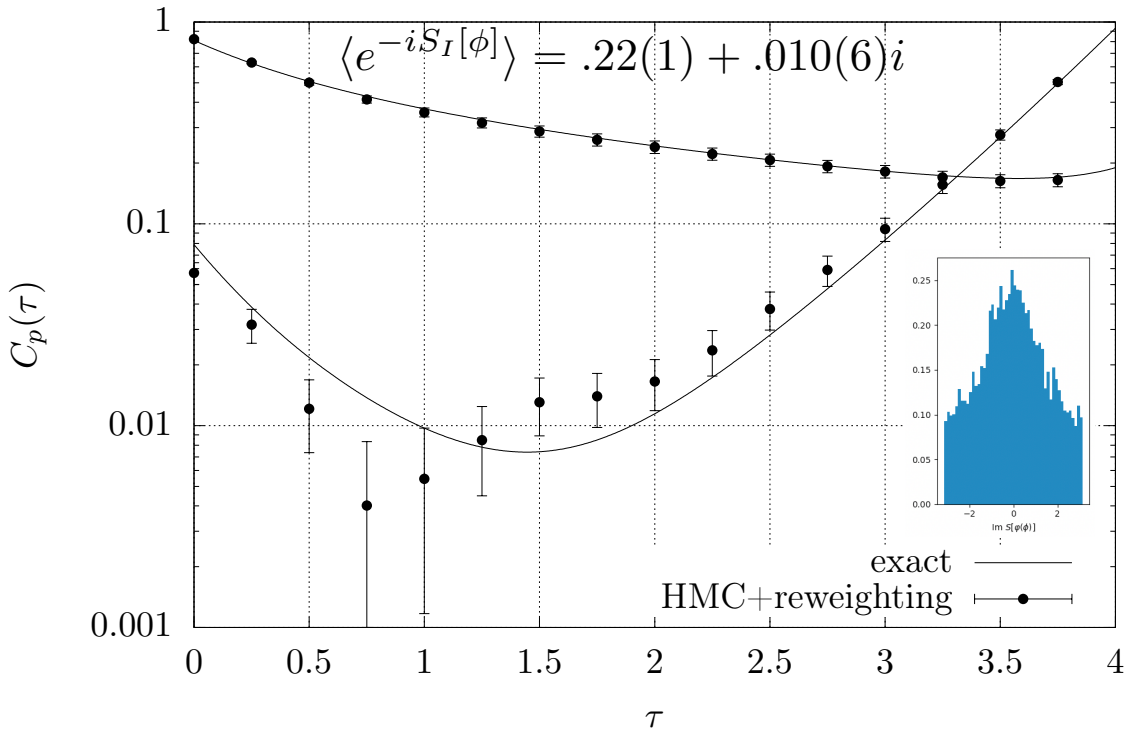


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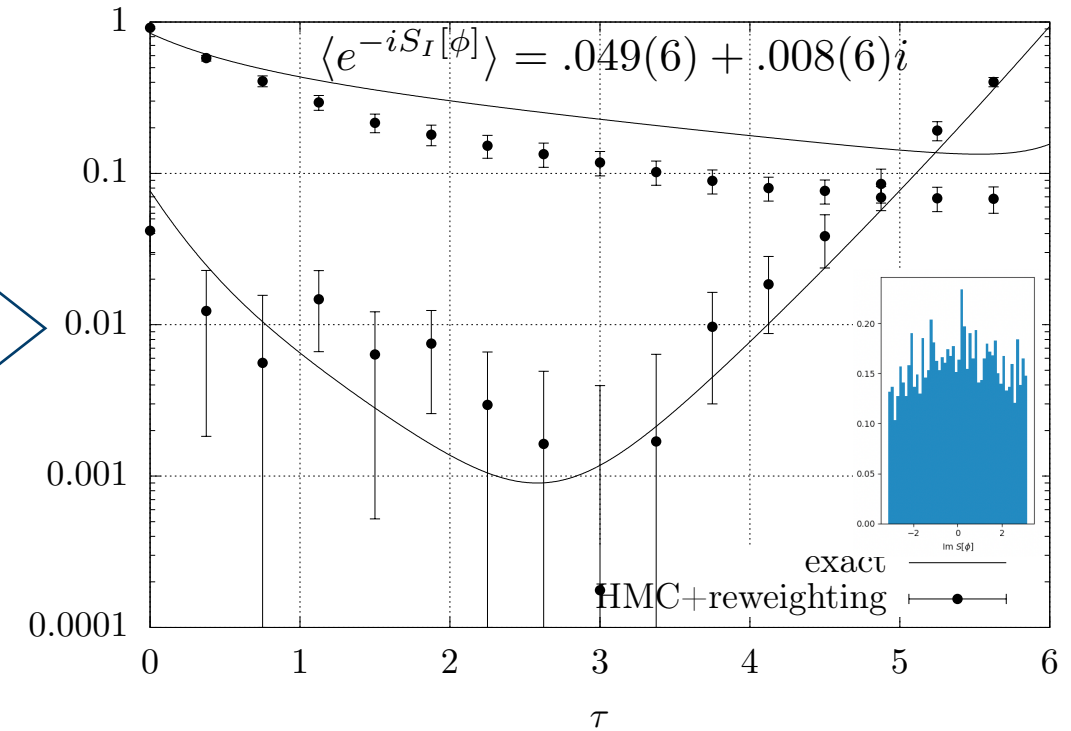
Three-site problem

$$U = 3, \beta = 4, N_t = 16$$

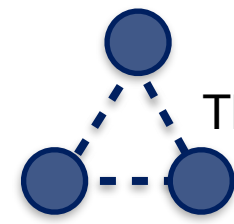


Increase
system size

$$U = 3, \beta = 6, N_t = 16$$

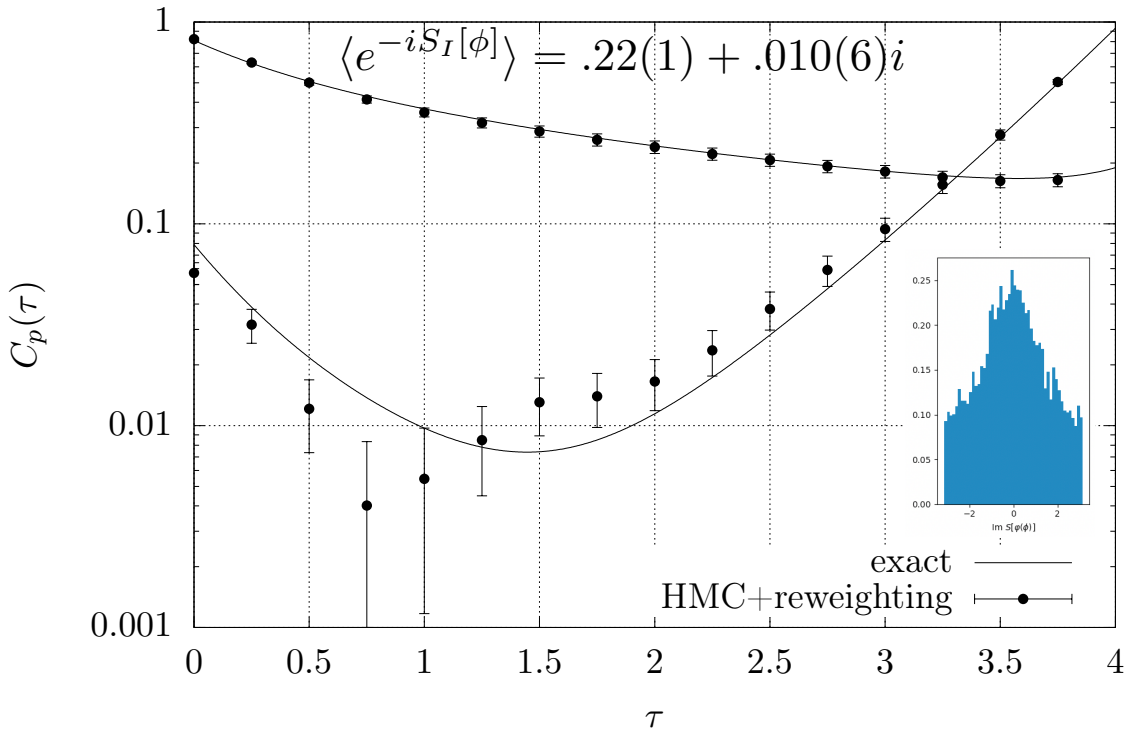


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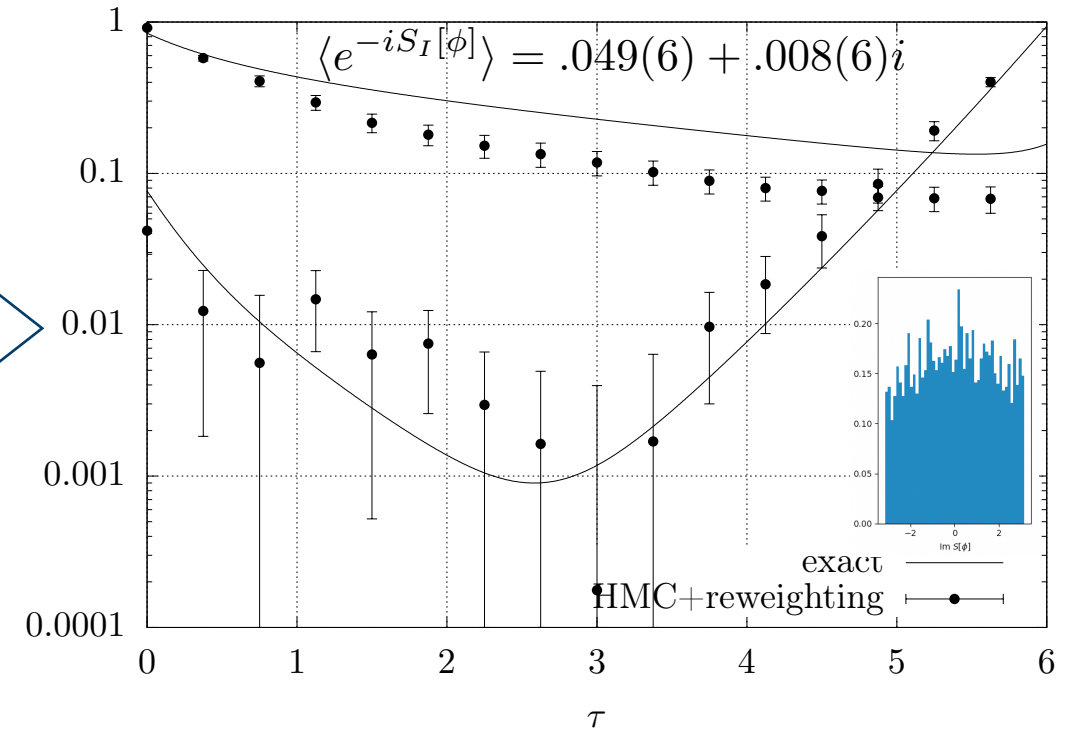
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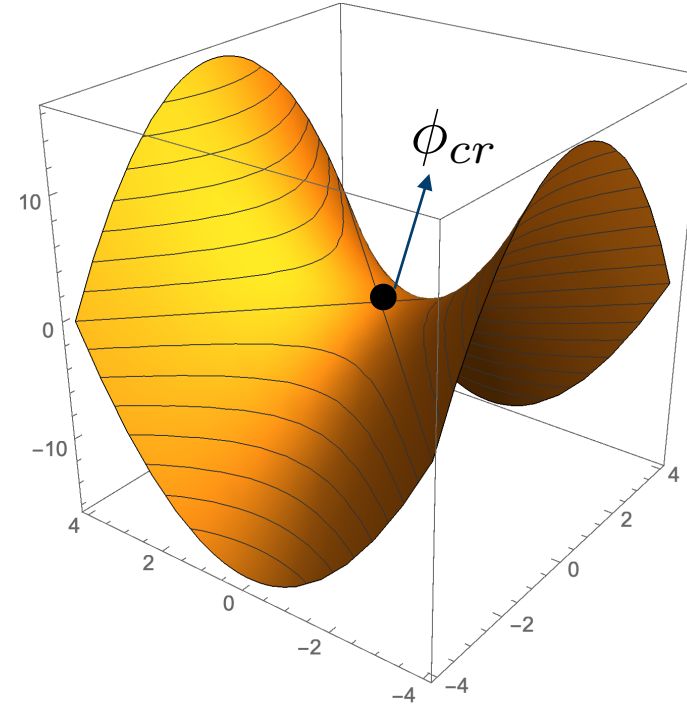


Phase fluctuations scale with the system size (extensive quantity). The larger the system, the worse the fluctuations $\Rightarrow$ re-weighting will inevitably fail for large enough system sizes

ALLEVIATING THE SIGN-PROBLEM VIA HOLOMORPHIC FLOW

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$$\int \mathcal{D}[\phi] e^{-S[\phi]} = e^{-S[\phi_{cr}]} \int \mathcal{D}[\phi] e^{-\delta S[\phi]}$$



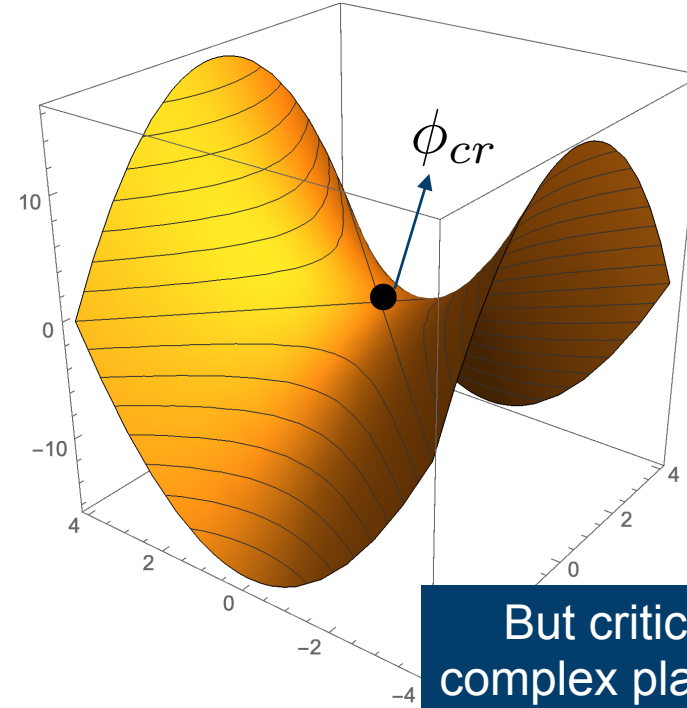
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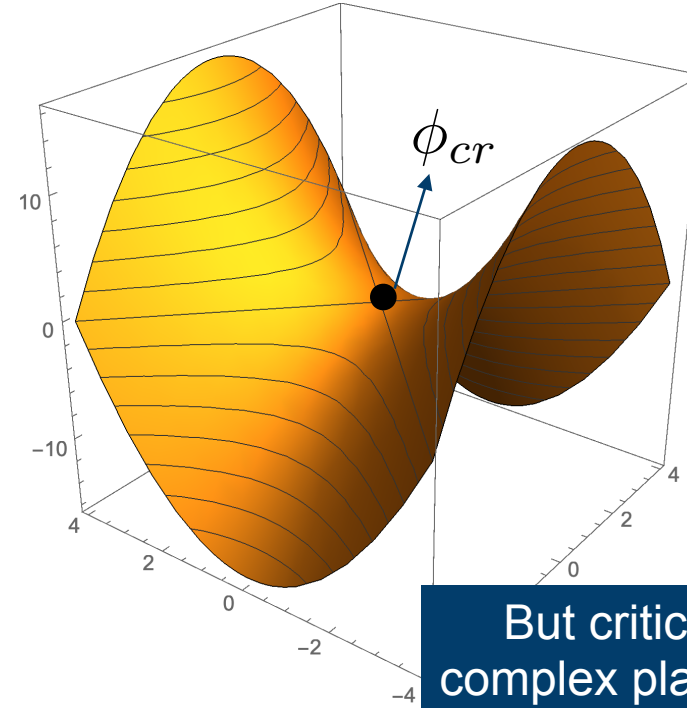
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 - Cauchy's theorem tells us our integral will remain the same as long as we don't cross any singularities
 - Preserve homology class of kernel

How does this help the sign problem?

- $\exists$ particular manifolds in the complex plane that **intersect** ϕ_{cr} and have $S_I[\phi]$ is constant!
 - Lefschetz thimbles



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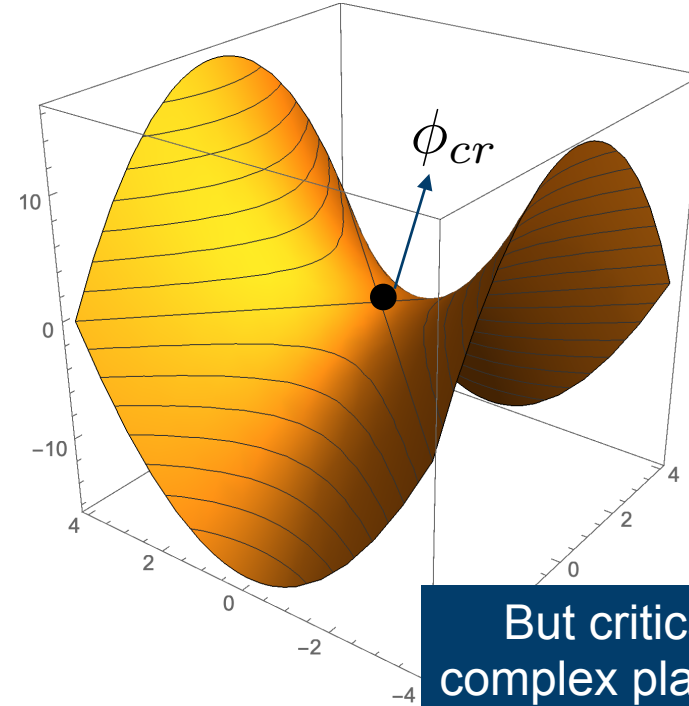
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$$\int_{\phi \in \mathbb{R}^N} \mathcal{D}[\phi] e^{-S[\phi]} \rightarrow e^{-iS_I^T} \int_{\varphi \in \mathcal{T} \subset \mathbb{C}^N} D[\varphi] e^{-S_R[\varphi]}$$



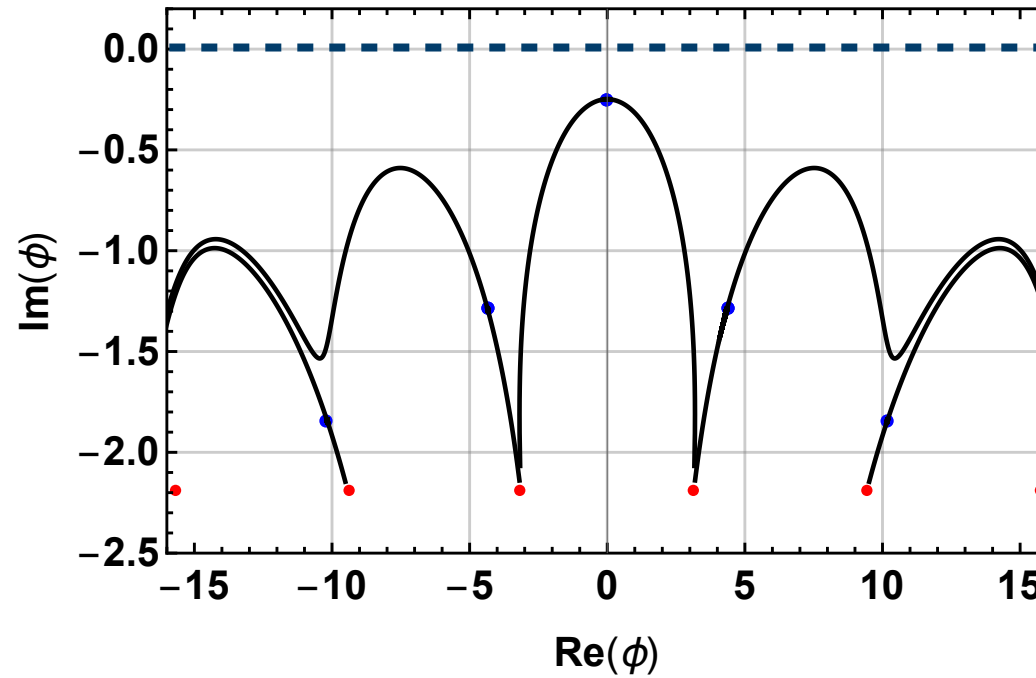
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MORE ON LEFSCHETZ THIMBLES

1d example: $S(\phi) = \frac{1}{20} \left(\phi + \frac{3i}{2} \right)^2 - \log \left(1 + \frac{1}{2} e^{i(\phi + \frac{3i}{2})} \right)$

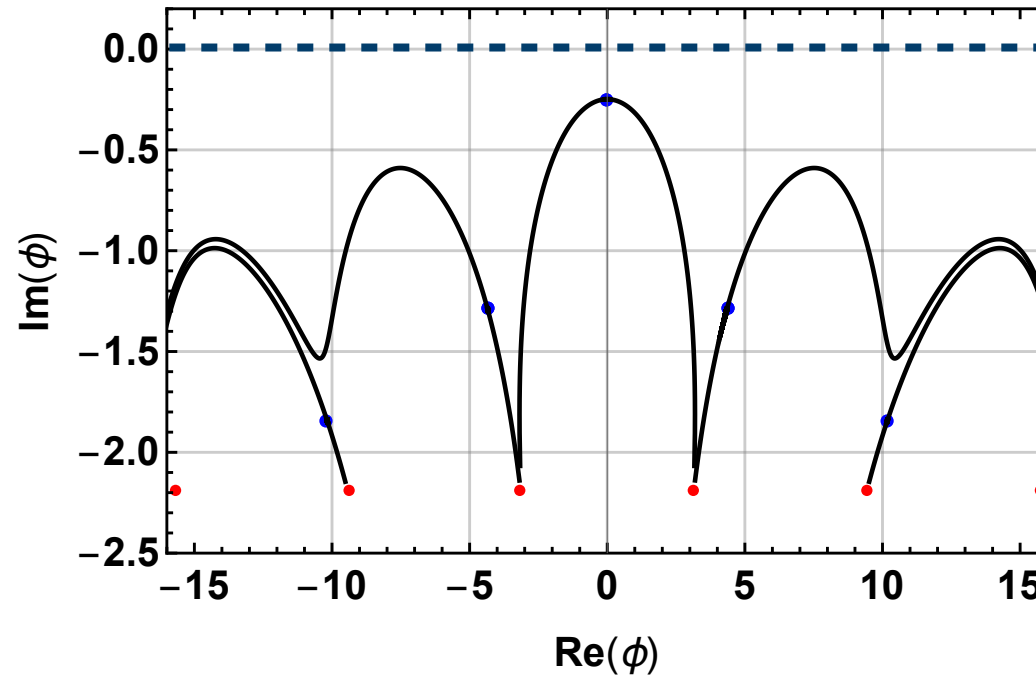


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Each critical point has an associated thimble

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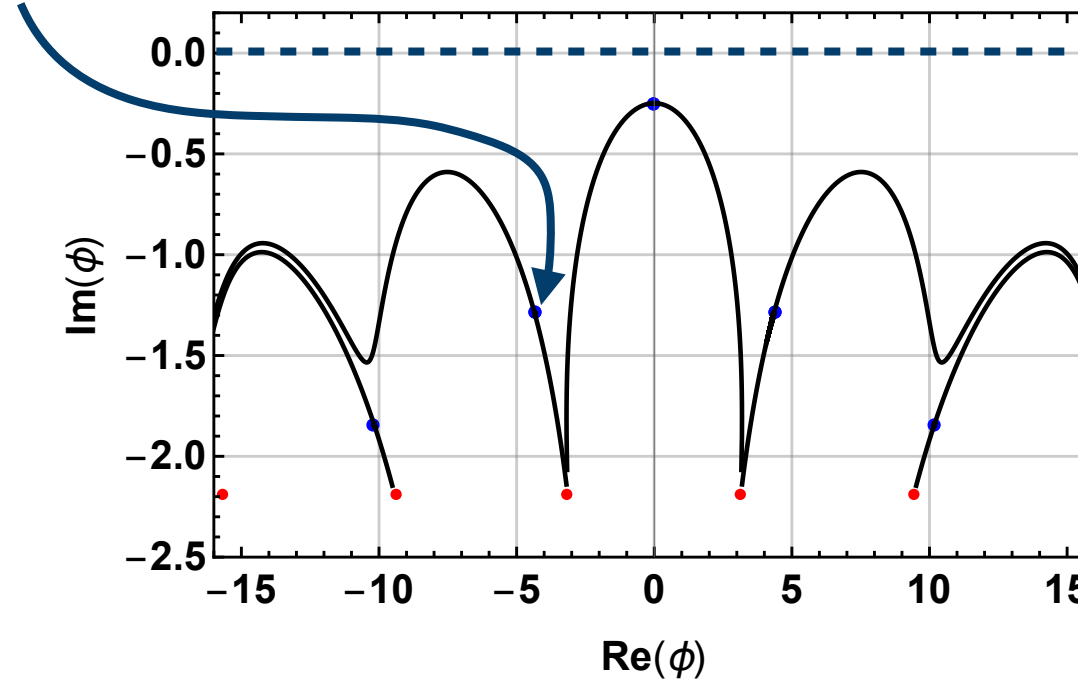


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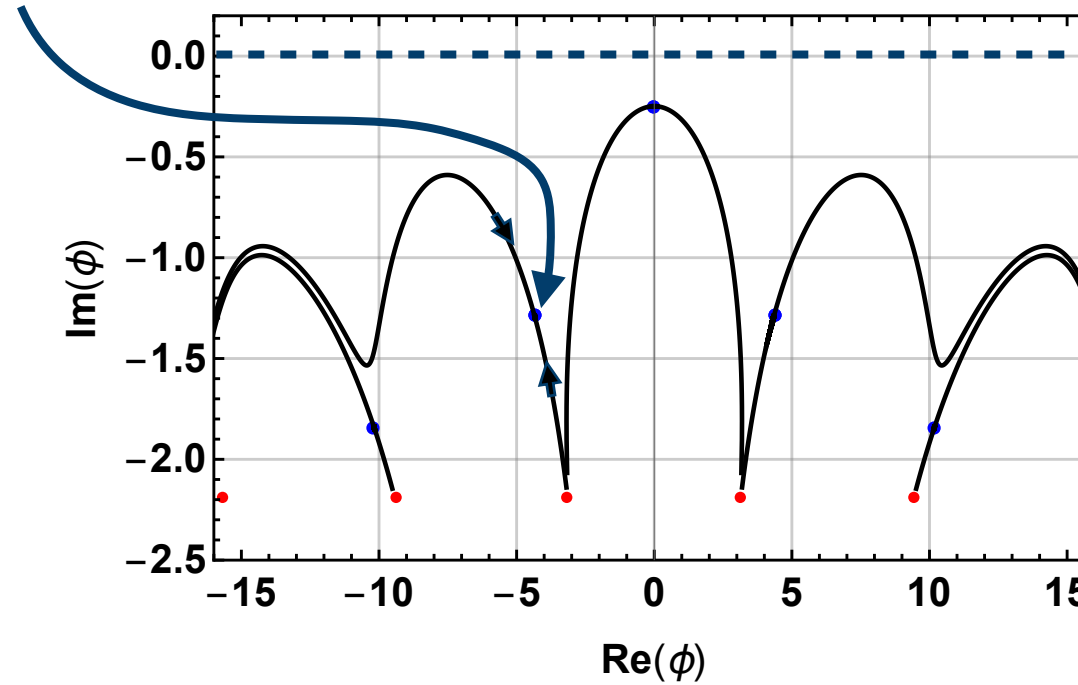
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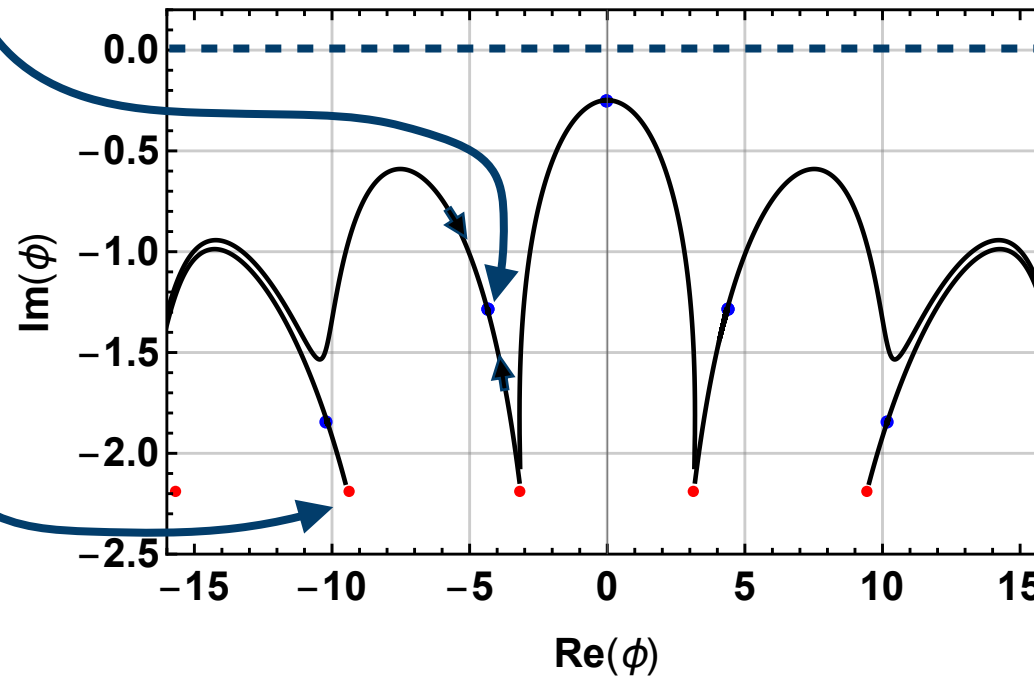
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Thimbles extend to infinity, or terminate at zeros

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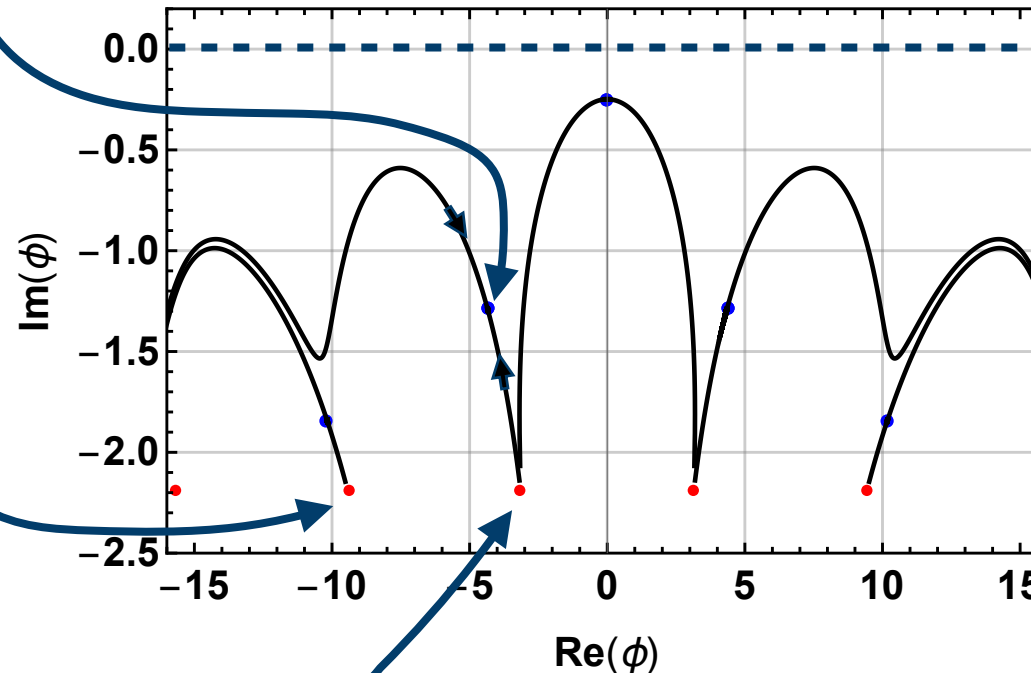
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$$\dot{\phi} = -(\nabla_{\phi} S[\phi])^*$$

Thimbles extend to infinity, or terminate at zeros

Thimbles don't cross each other, but they can connect via the zeros

1d example: $S(\phi) = \frac{1}{20} \left(\phi + \frac{3i}{2} \right)^2 - \log \left(1 + \frac{1}{2} e^{i(\phi + \frac{3i}{2})} \right)$



MORE ON LEFSCHETZ THIMBLES

Each critical point has an associated thimble

$$\mathcal{J}(\phi_{cr}) = \{\phi \mid \lim_{t \rightarrow \infty} \phi(t) = \phi_{cr}\}$$

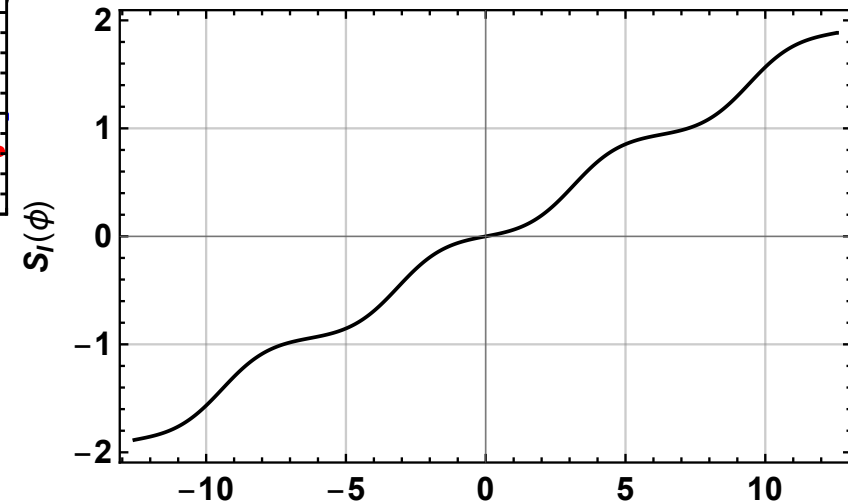
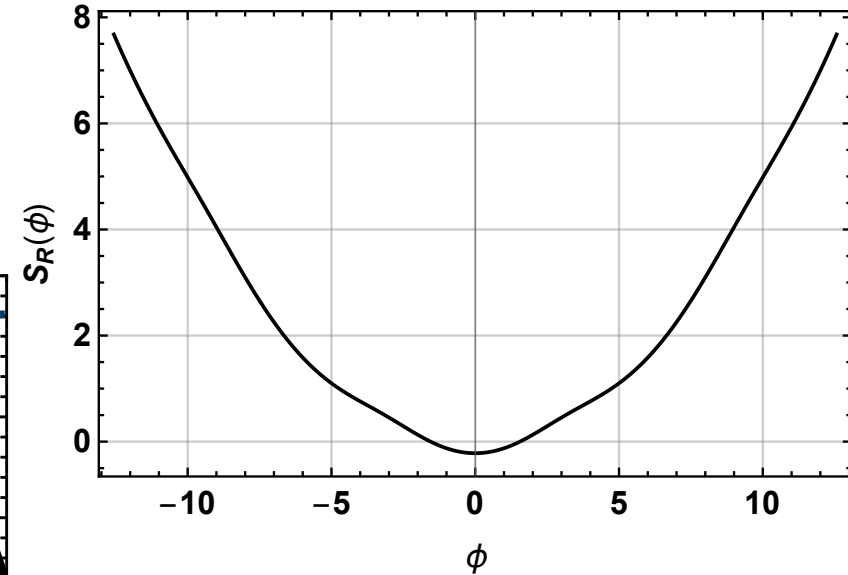
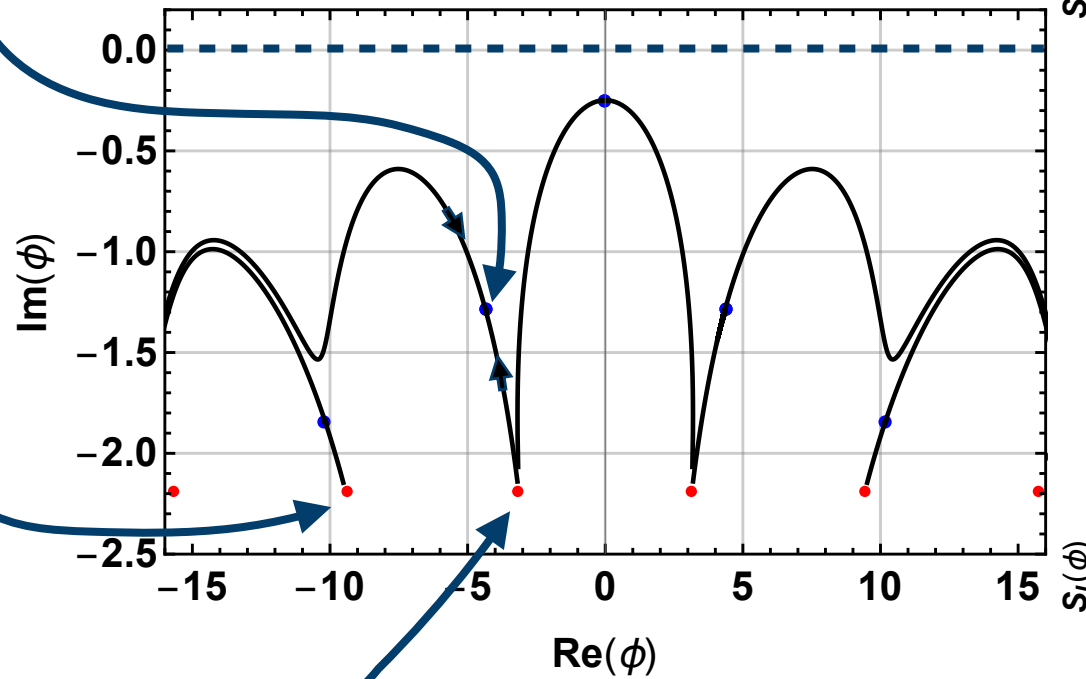
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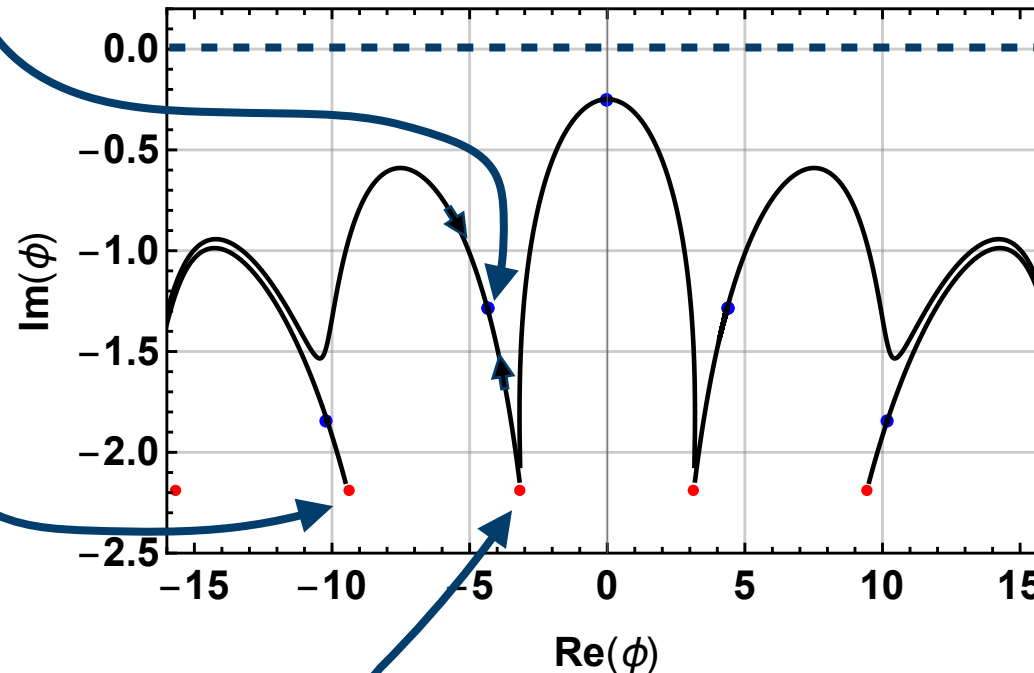
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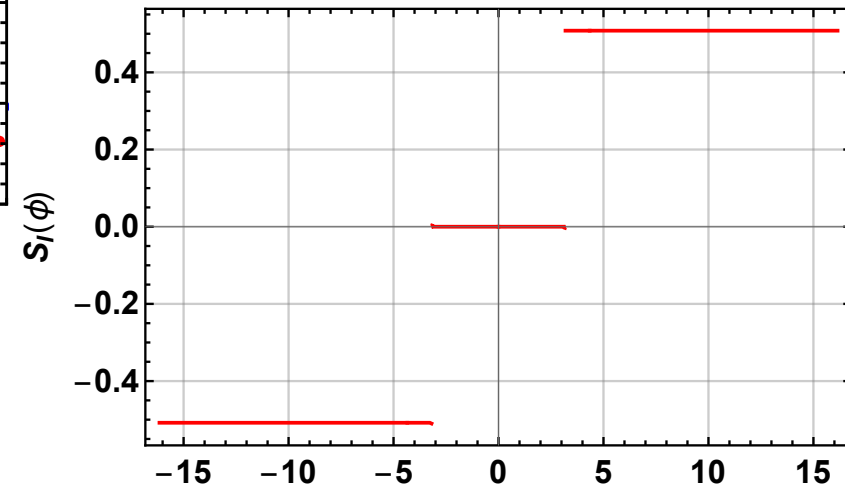
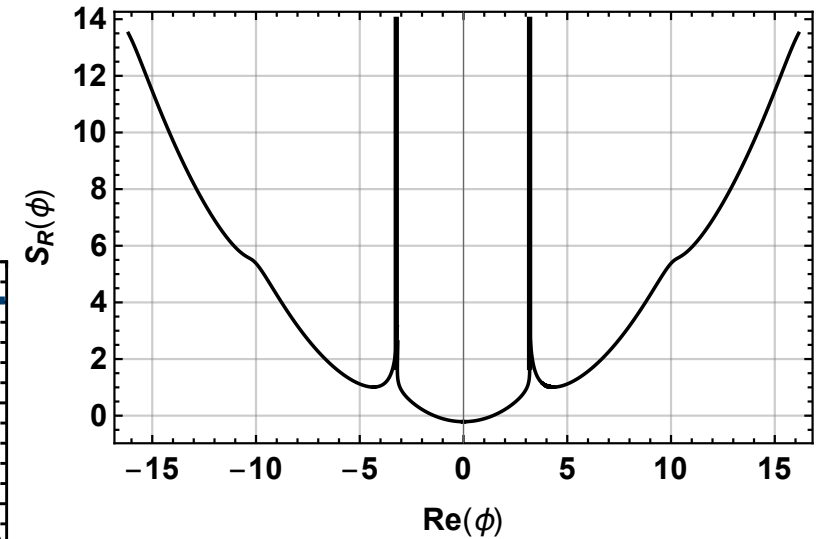
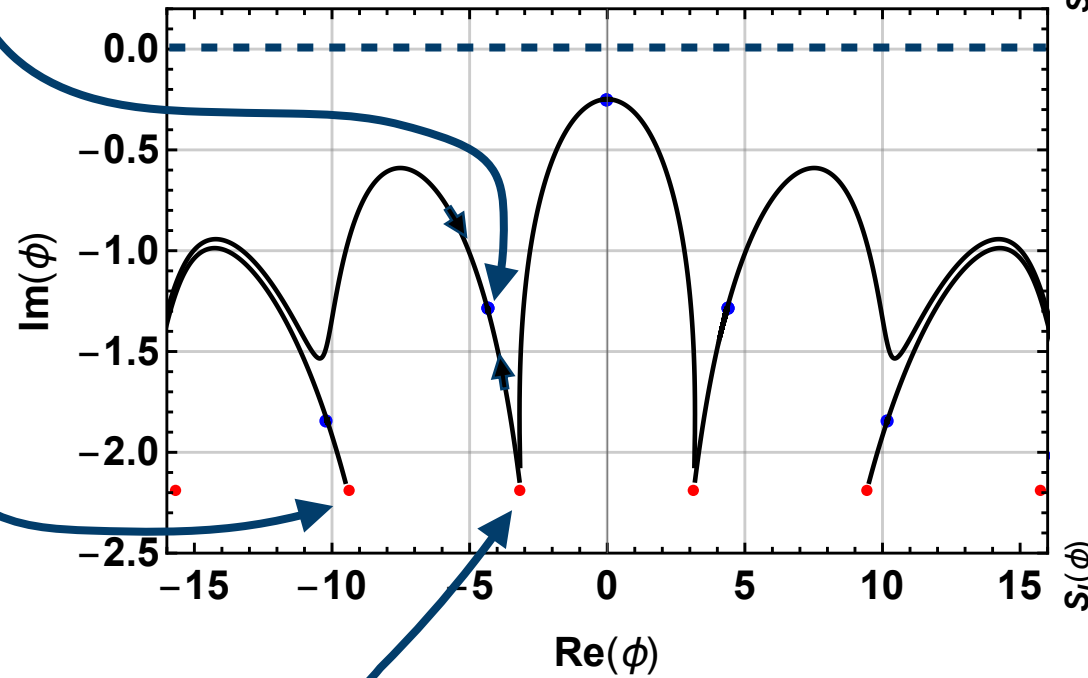
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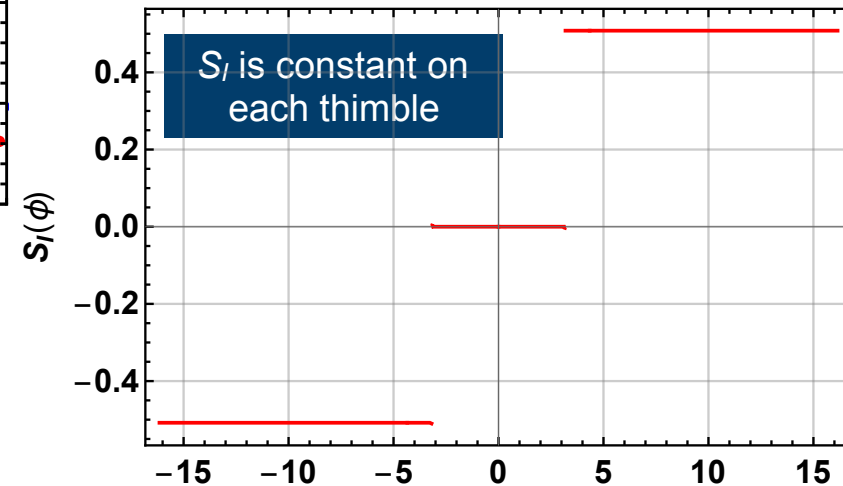
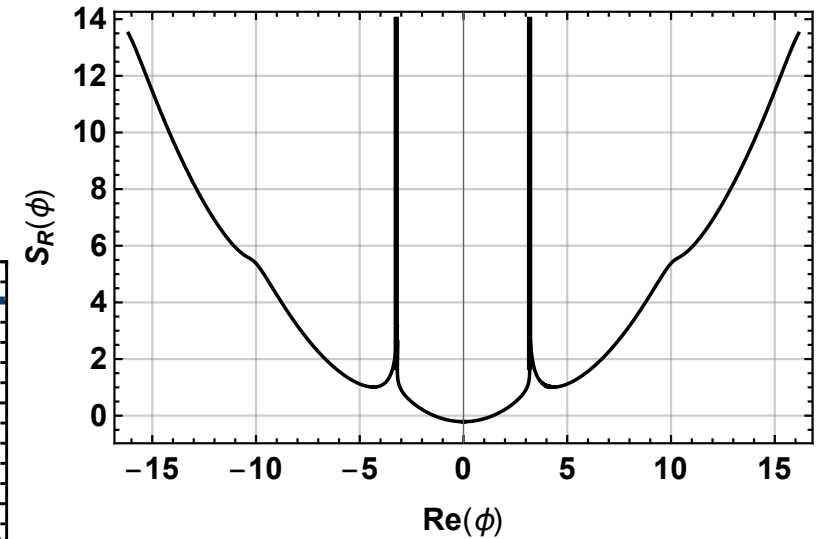
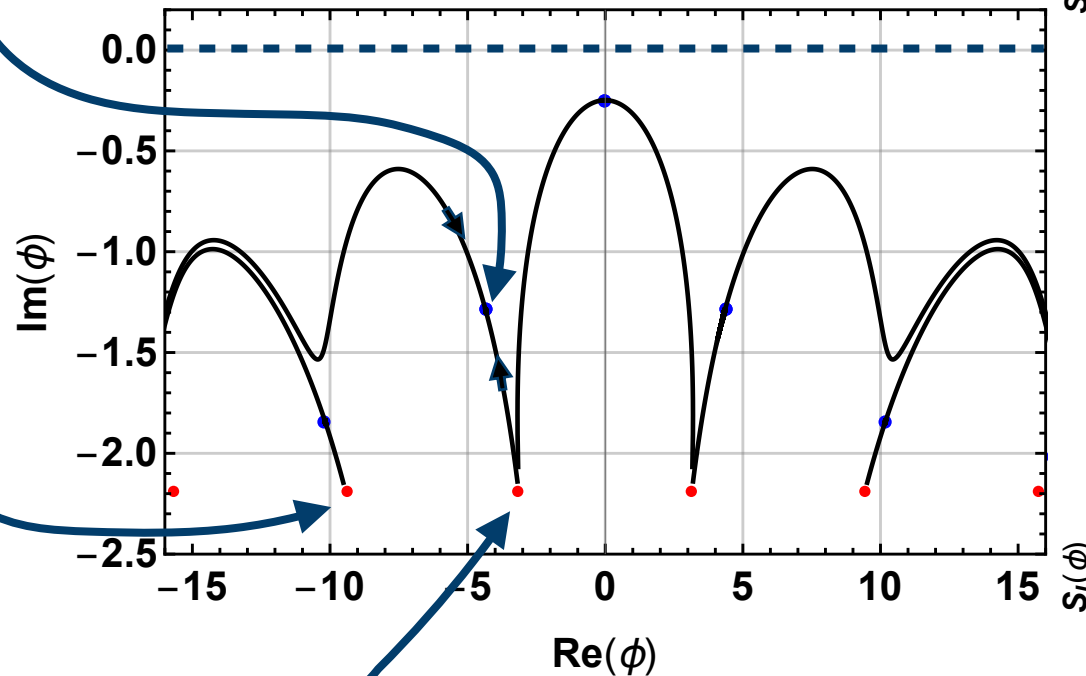
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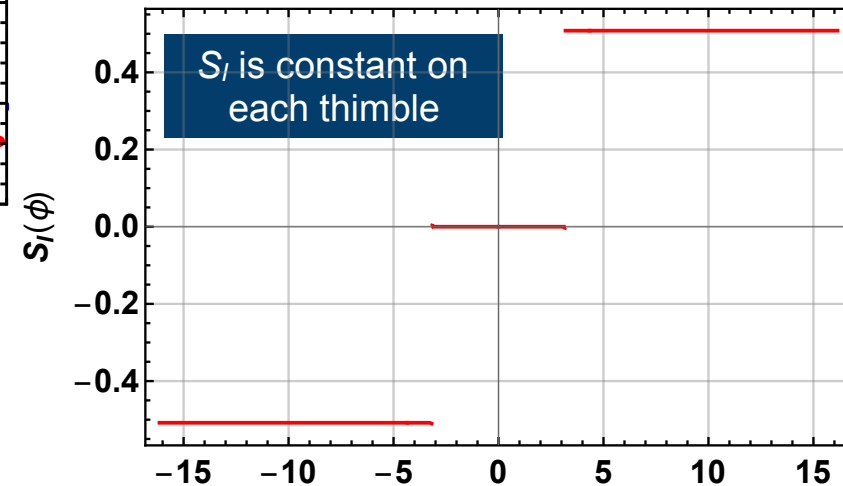
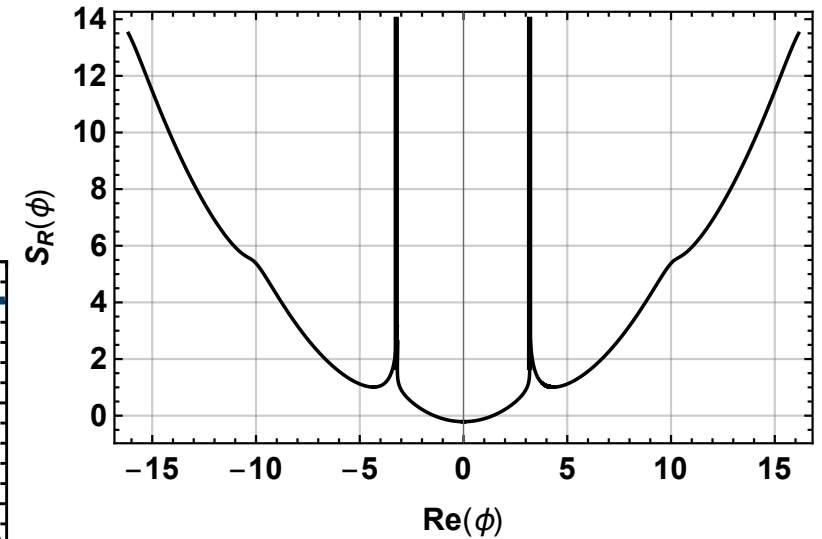
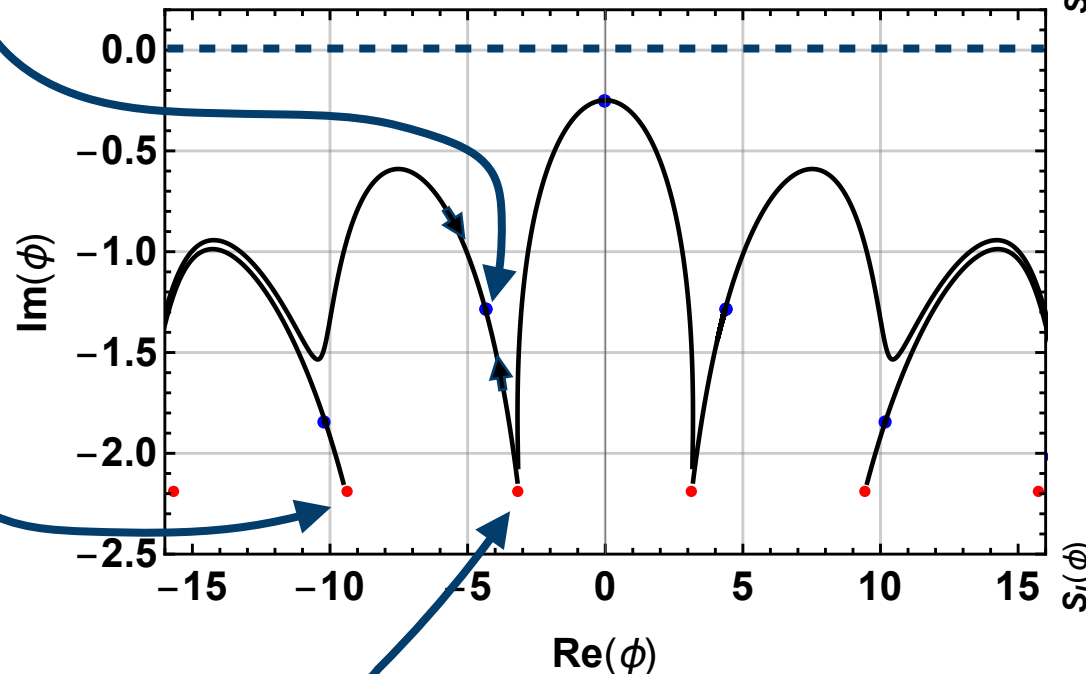
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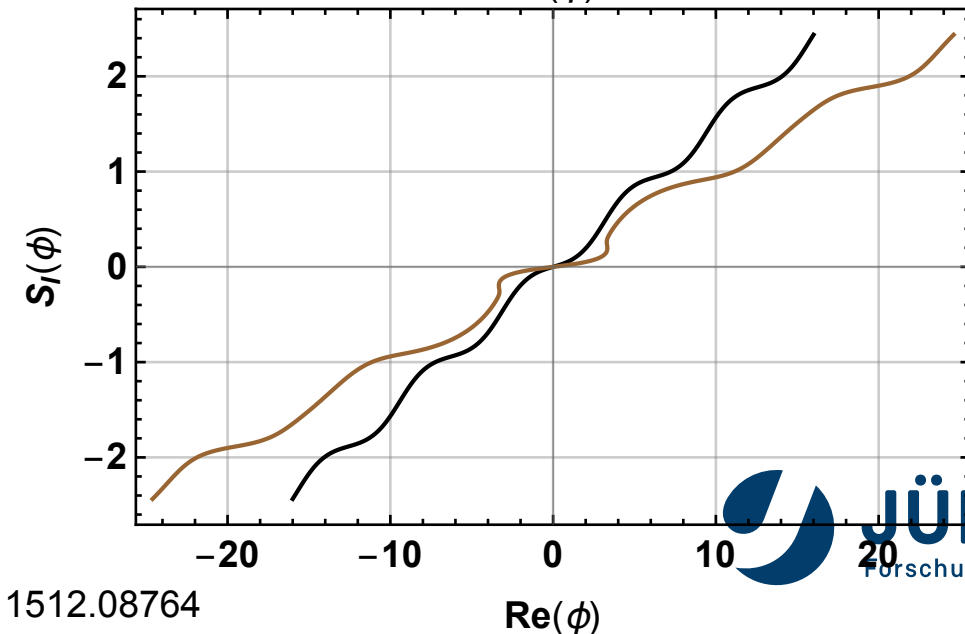
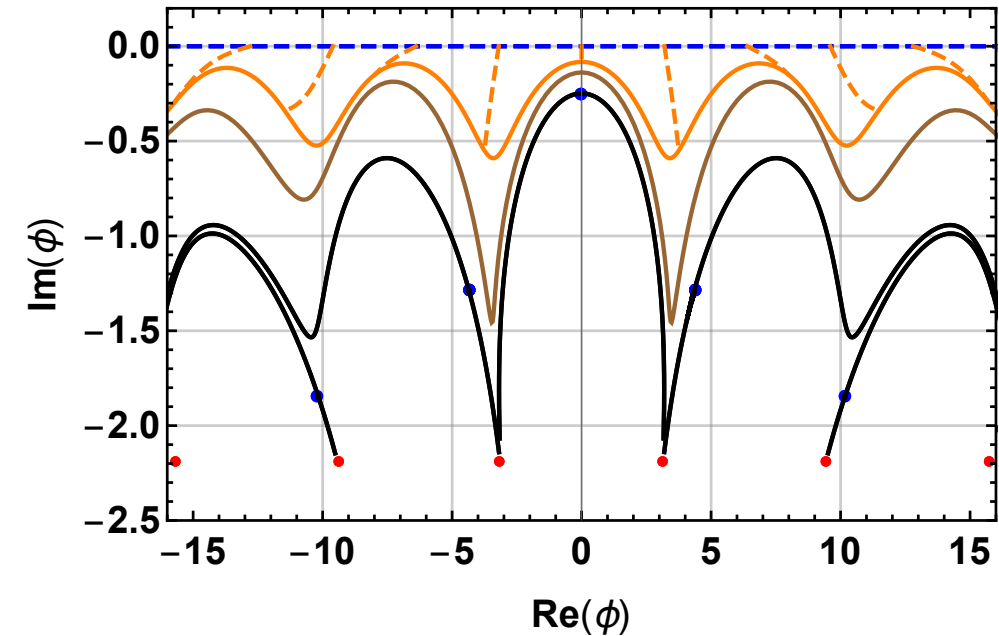
But we don't know *a priori* where these thimbles are!

BUT WE CAN APPROXIMATE THIMBLES BY FLOWING FROM REAL PLANE

$$\dot{\phi} = (\nabla_{\phi} S[\phi])^*$$

- Flow equations respect the homology class
- Starting from any point on the real plane, the flow will approach some point on the thimble manifolds
 - In the limit of infinite flow time, one recovers the exact thimbles (care must be taken in presence zeros)
- For finite flow time, one has an approximation to the thimbles
 - Sign problem has been alleviated
 - Re-weighting should work!

“Generalized Thimble” approach



HOW DO WE SAMPLE FROM THE MANIFOLDS?

Integration on manifolds



Integration on real plane

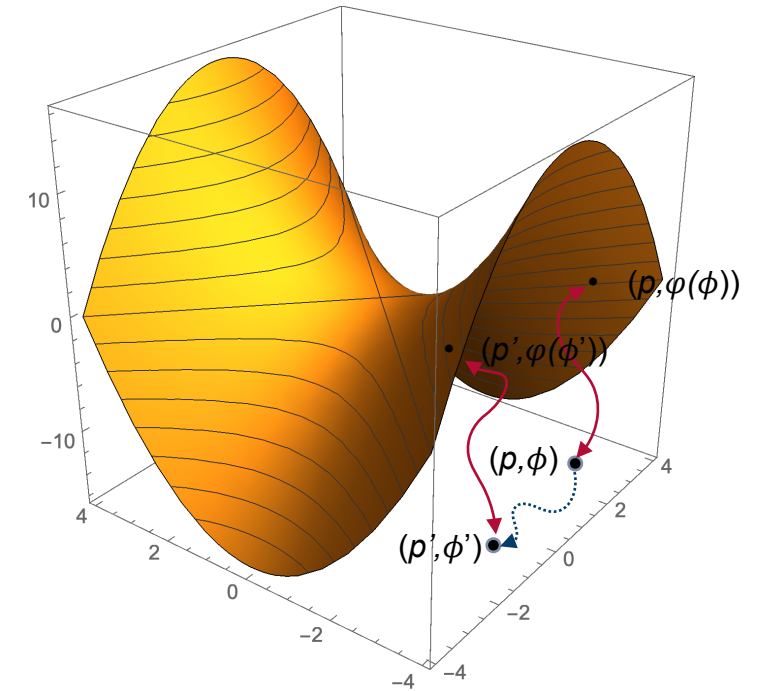
$$\begin{aligned} \frac{1}{\mathcal{Z}_{\mathcal{T}}} \int_{\mathcal{T}} D[\varphi] O[\varphi] \exp \{-S[\varphi]\} &= \frac{1}{\mathcal{Z}_{\mathcal{T}}} \int_{\mathbb{R}} D[\phi] O[\varphi(\phi)] \exp \{-S[\varphi(\phi)]\} \det J[\varphi(\phi)] \\ &= \frac{1}{\mathcal{Z}_{\mathcal{T}}} \int_{\mathbb{R}} D[\phi] O[\varphi(\phi)] \exp \{-S_{eff}[\varphi(\phi)]\} \end{aligned}$$

$$S_{eff}[\varphi(\phi)] \equiv S[\varphi(\phi)] - \log \det J[\varphi(\phi)]$$

Jacobian is induced due to the change of variables from the manifold to the real plane

HMC WITH HOLOMORPHIC FLOW

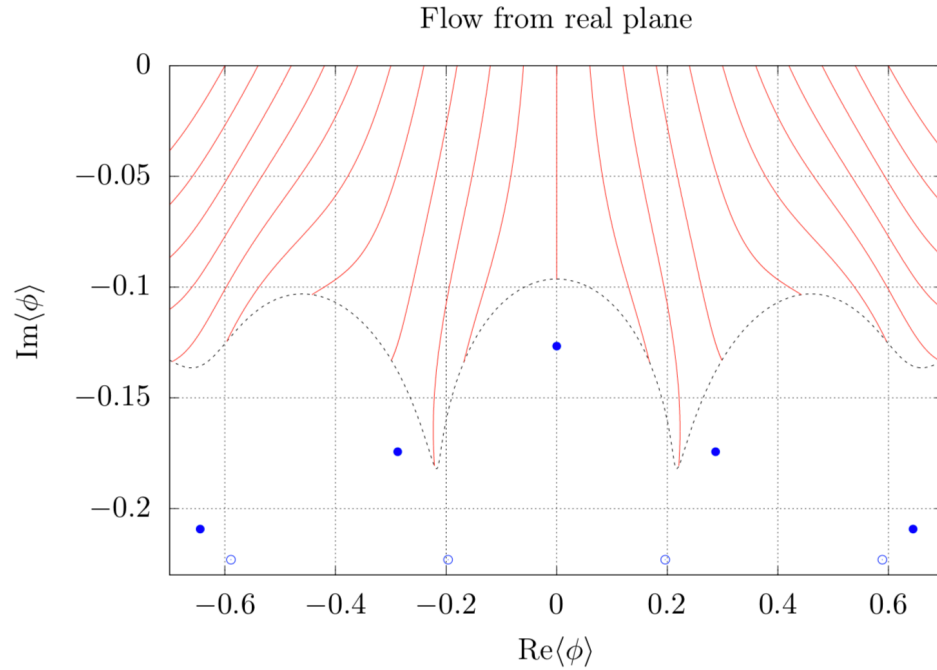
- Perform HMC on the real plane to make proposal $(p, \phi) \rightarrow (p', \phi')$
- Then flow up (close) to manifold $\varphi(\phi')$
- Accept/reject with $\mathbb{P}_{a/r} = \min \left(1, \frac{e^{-p'^2/2 - \text{Re}S_{eff}[\varphi(\phi')]} }{e^{-p^2/2 - \text{Re}S_{eff}[\varphi(\phi)]}} \right)$
- Reversible, satisfies detailed balance



HMC directly on the thimble:
Ulybyshev et al. [[arXiv:1906.02726](https://arxiv.org/abs/1906.02726)]

FLOWING THE ACTION

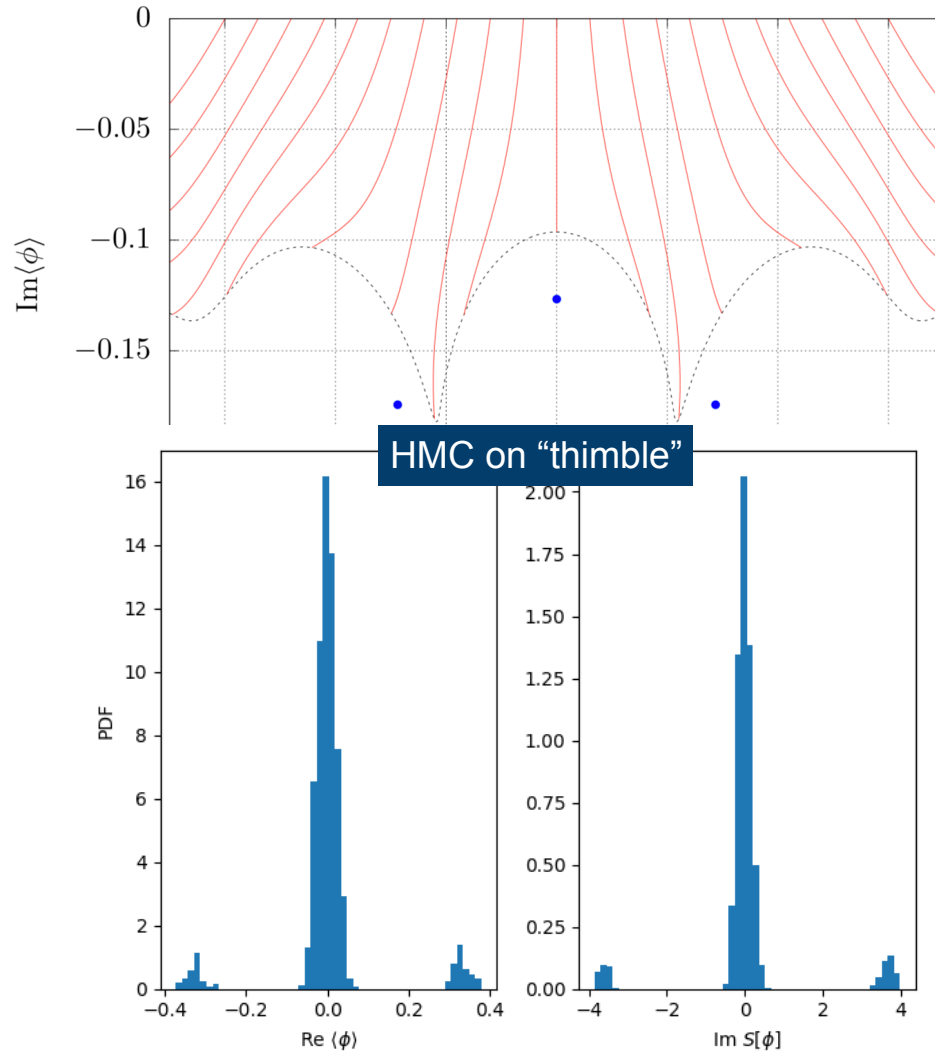
Flow from real plane $\dot{\phi} = (\nabla_{\phi} S[\phi])^*$



FLOWING THE ACTION

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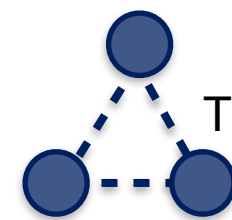
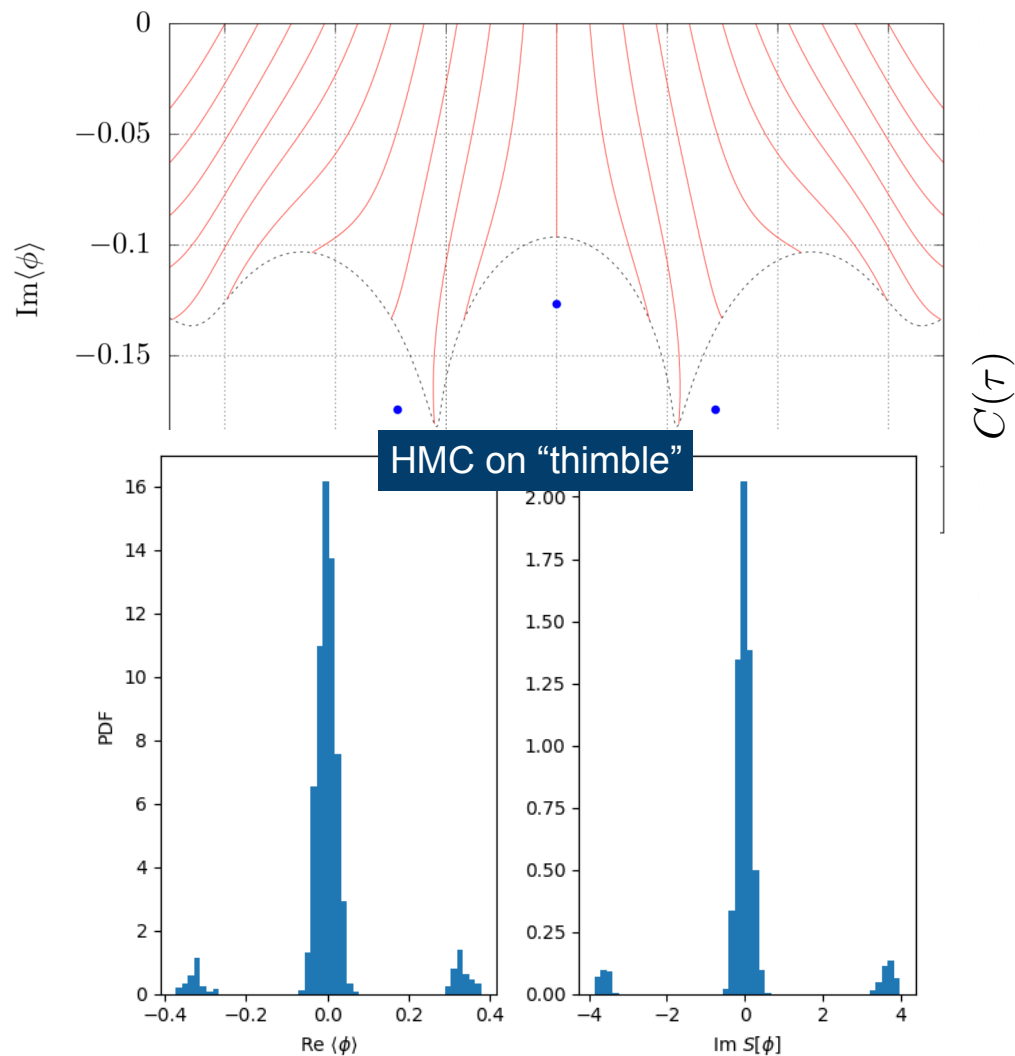
Flow from real plane



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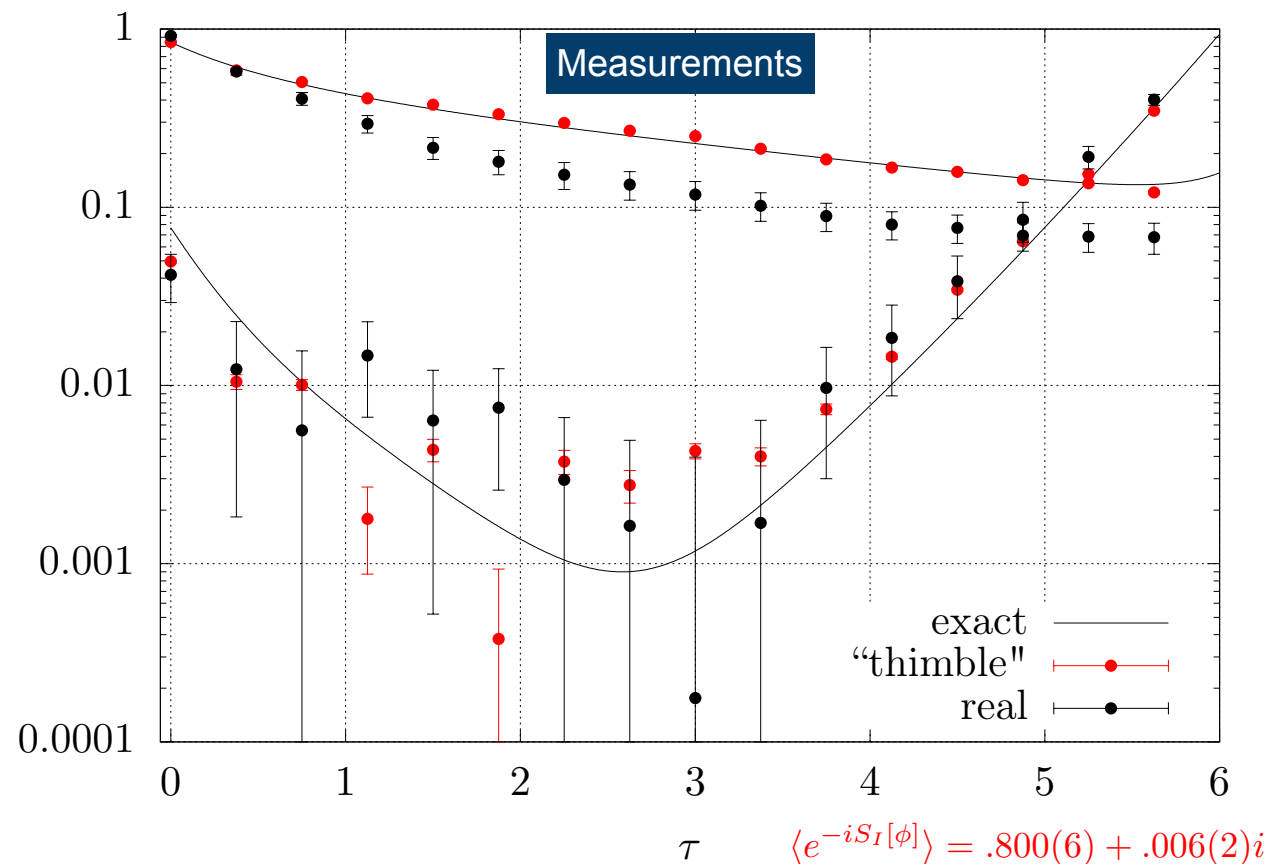
Flow from real plane $\dot{\phi} = (\nabla_{\phi} S[\phi])^*$

Flow from real plane



Three-site problem

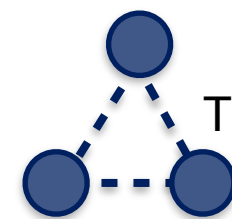
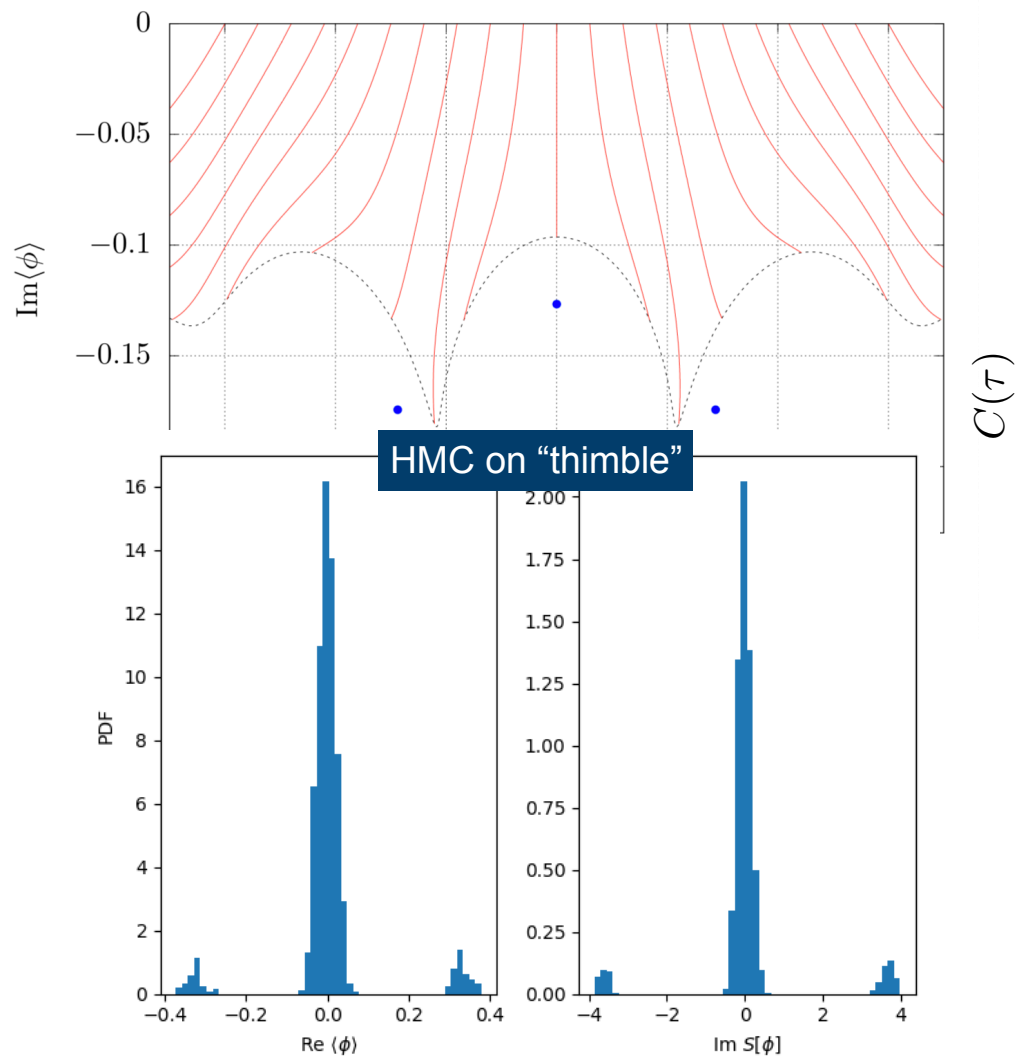
$$U = 3, \beta = 6, N_t = 16$$



FLOWING THE ACTION

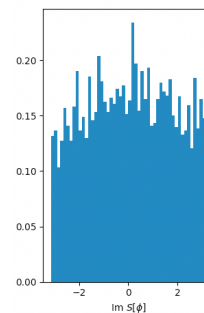
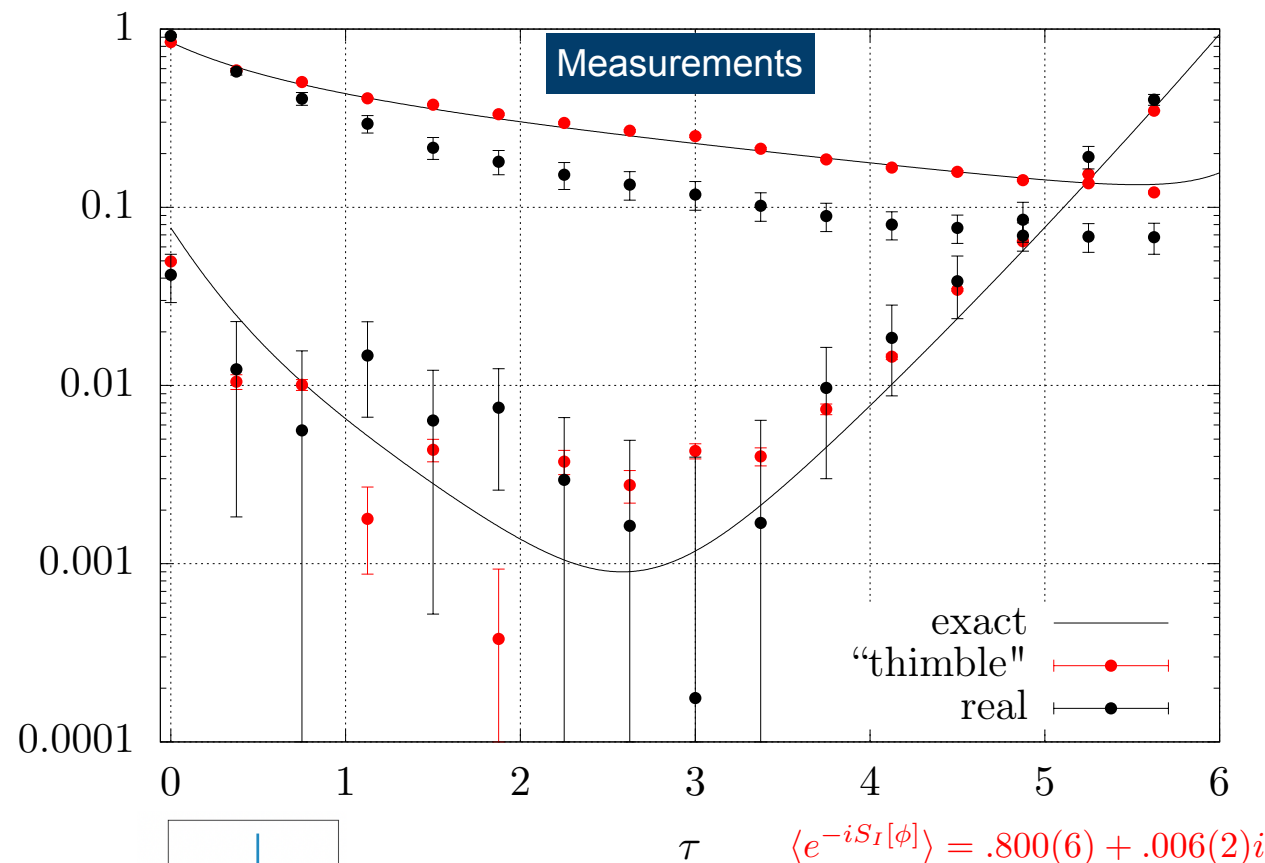
Flow from real plane $\dot{\phi} = (\nabla_{\phi} S[\phi])^*$

Flow from real plane



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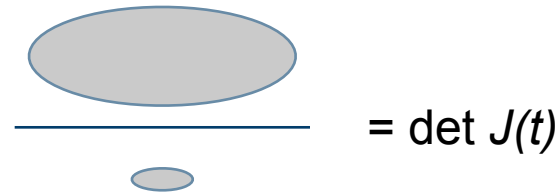
$$\langle e^{-iS_I[\phi]} \rangle = .049(6) + .008(6)i$$

UNFORTUNATELY, THIS “SOLUTION” IS NOT VIABLE NUMERICALLY

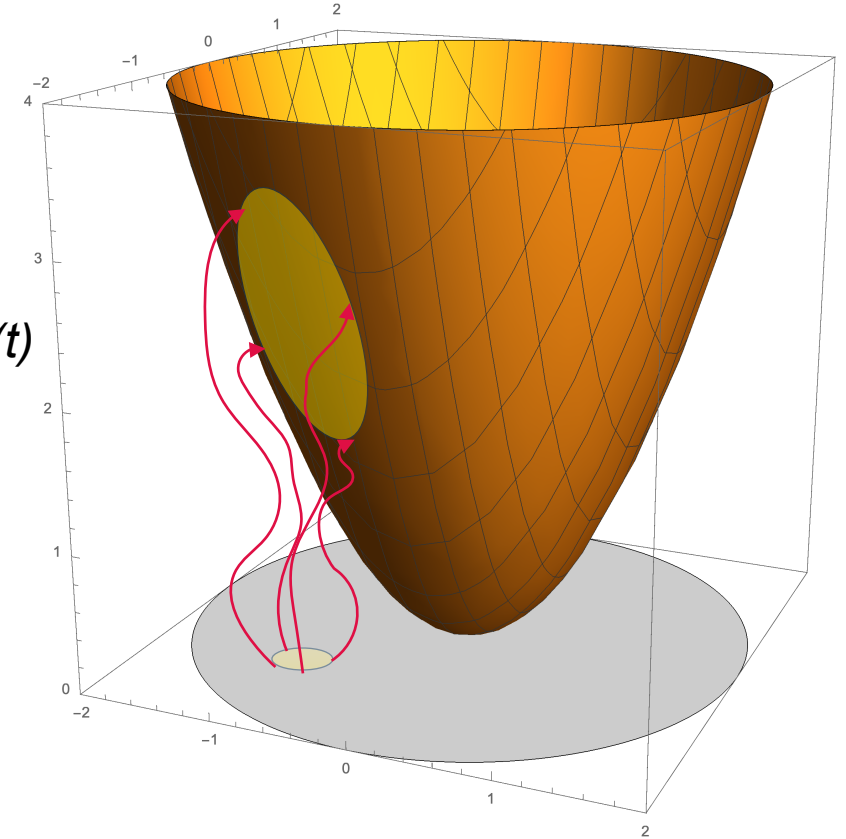
“No free lunch theorem”

$$\dot{\phi}(t) = [\nabla_{\phi} S[\phi(t)]]$$

$$\dot{J}(t) = [H[\phi(t)]J(t)]^*$$

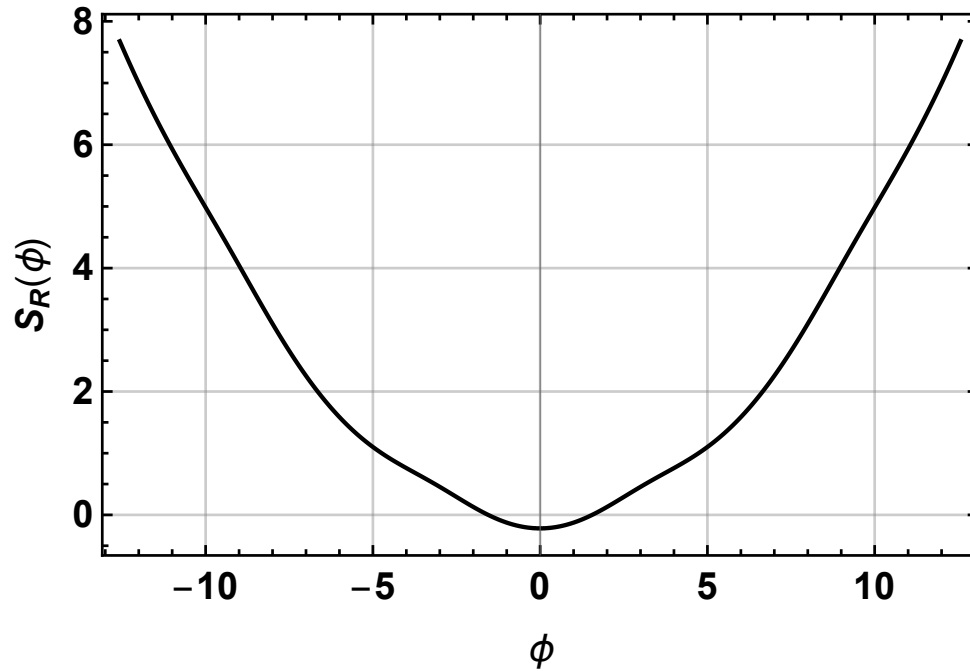


- Calculation of Jacobian due to flow is extremely time consuming!
- At every step of the flow integration, must compute a dense matrix!



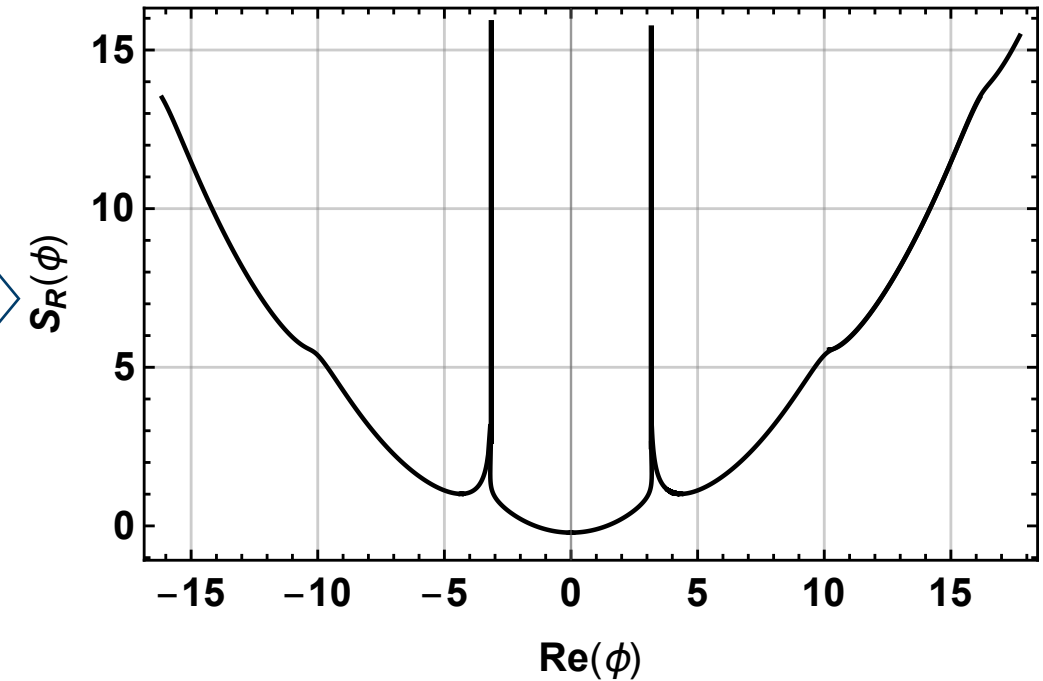
UNDER THE FLOW, MUST BE LEERY OF INFINITIES

no ergodicity issues, but sign problem



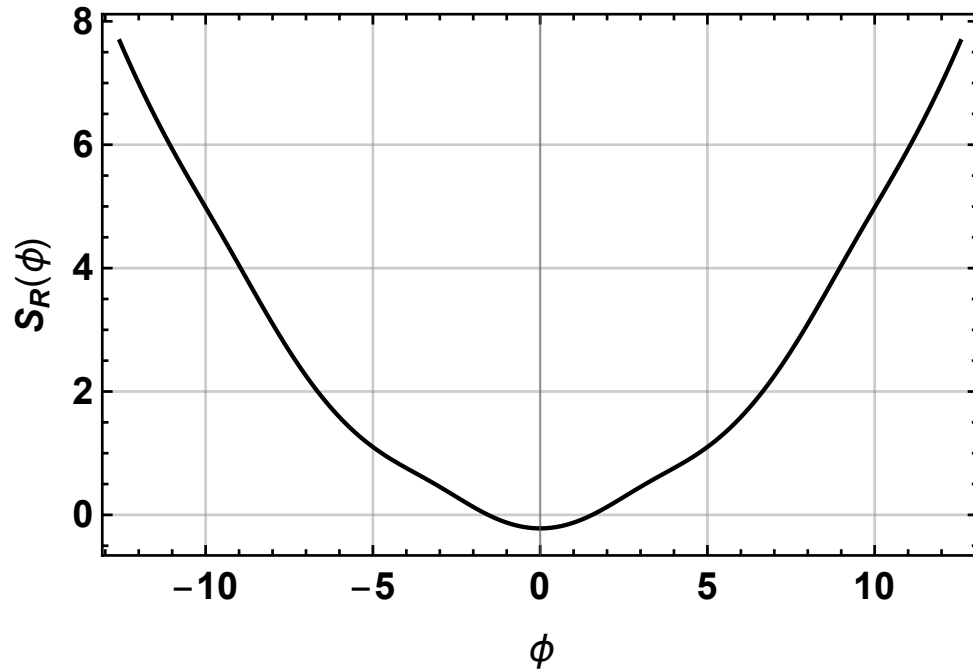
$$\dot{\phi} = (\nabla_{\phi} S[\phi])^*$$
$$\lim_{t_f \rightarrow \infty}$$

no sign problem, but ergodicity issues



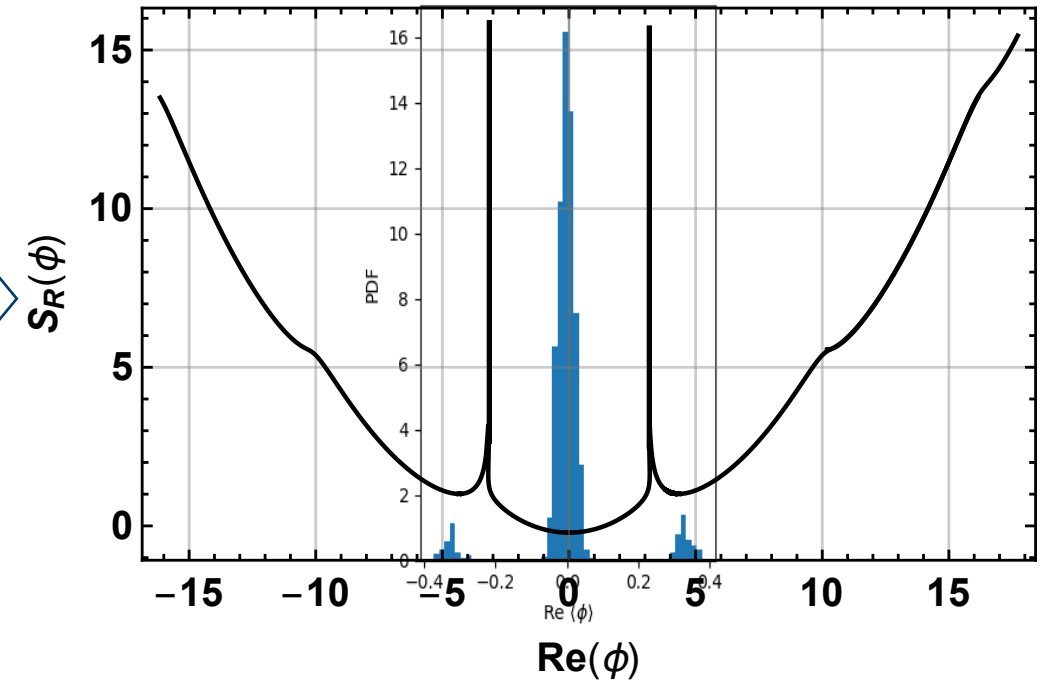
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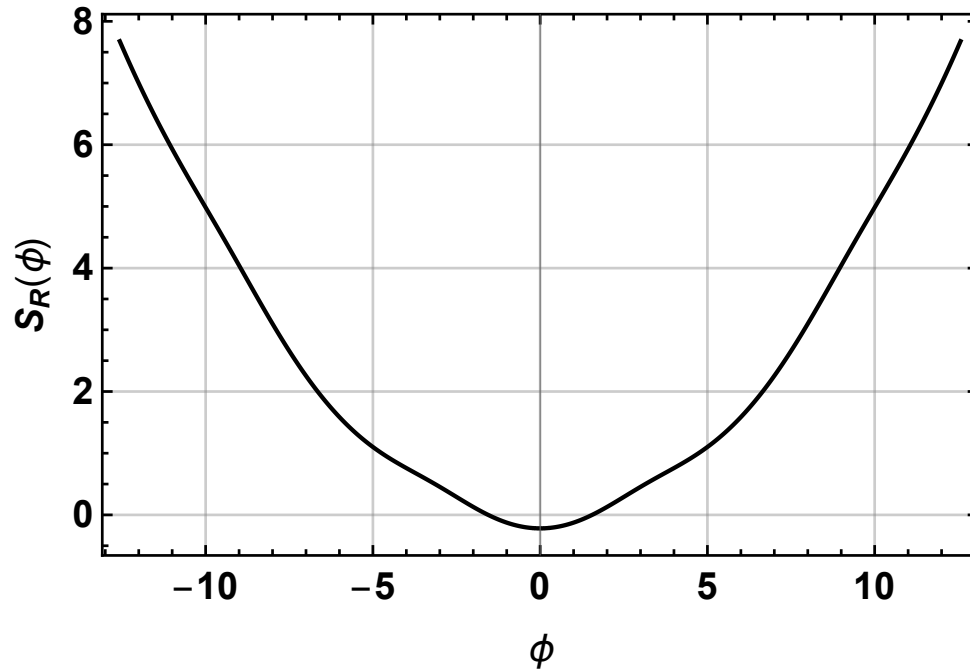
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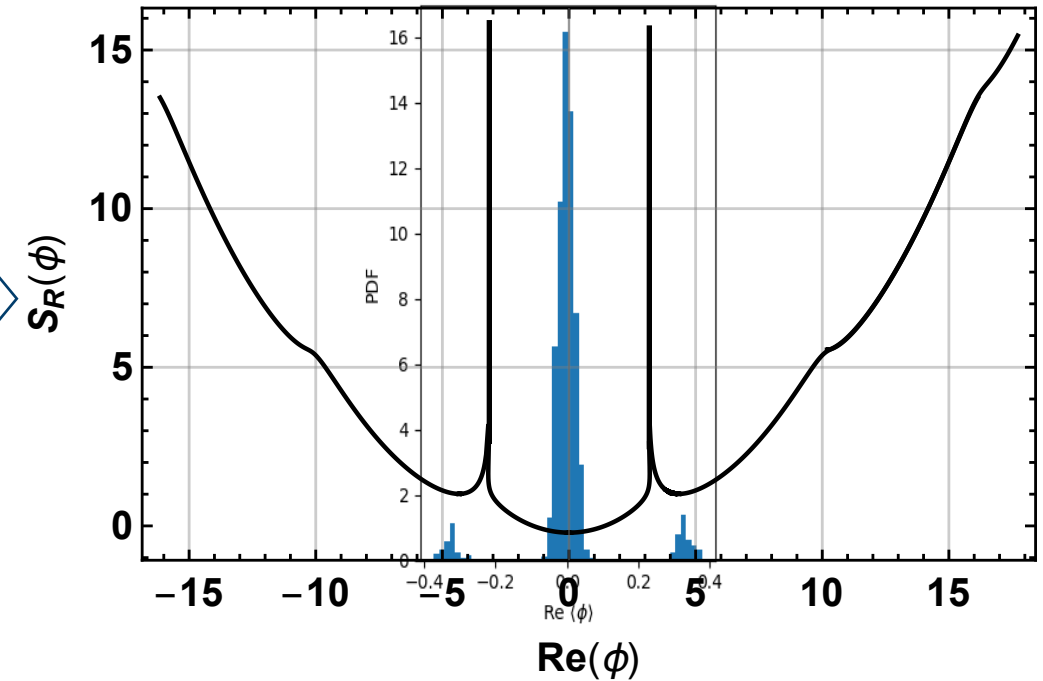
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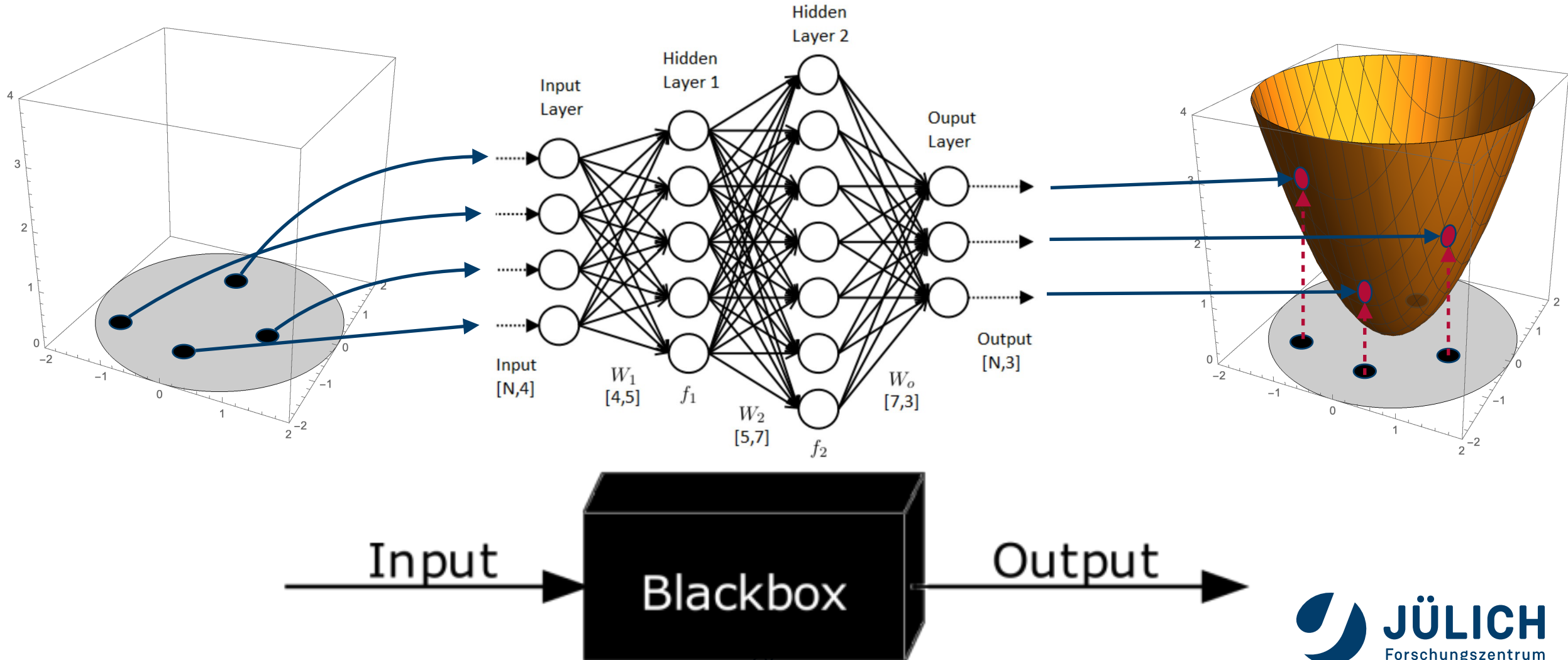
no sign problem, but ergodicity issues



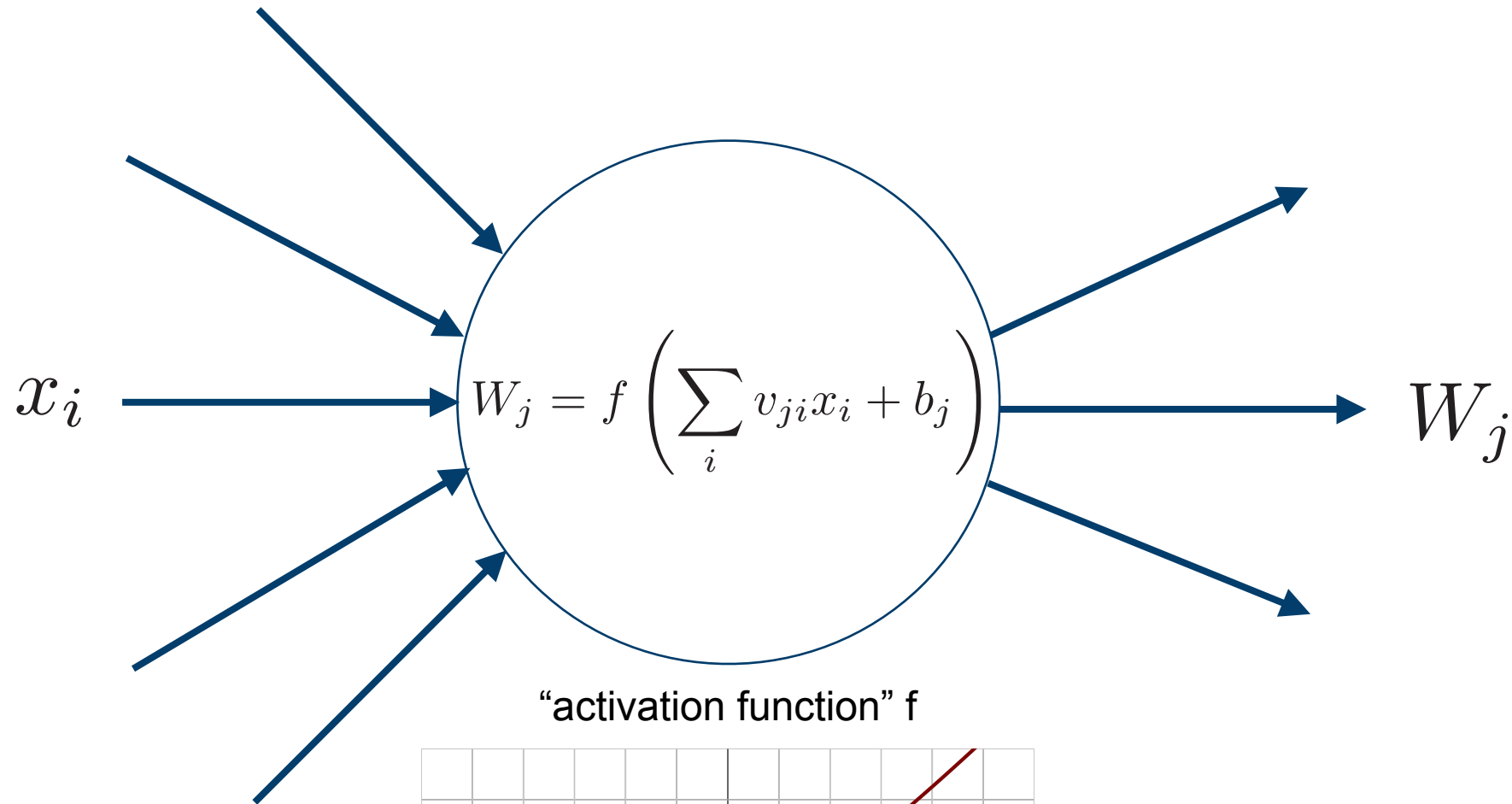
Distributions become highly multi-modal (ergodicity issues)

MACHINE LEARNING TO THE RESCUE!

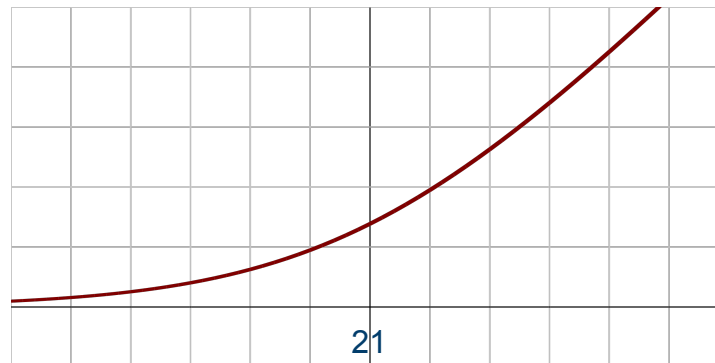
Pushing all our ignorance into one black box called a neural network (NN)



COMPOSITION OF A NEURONAL NODE



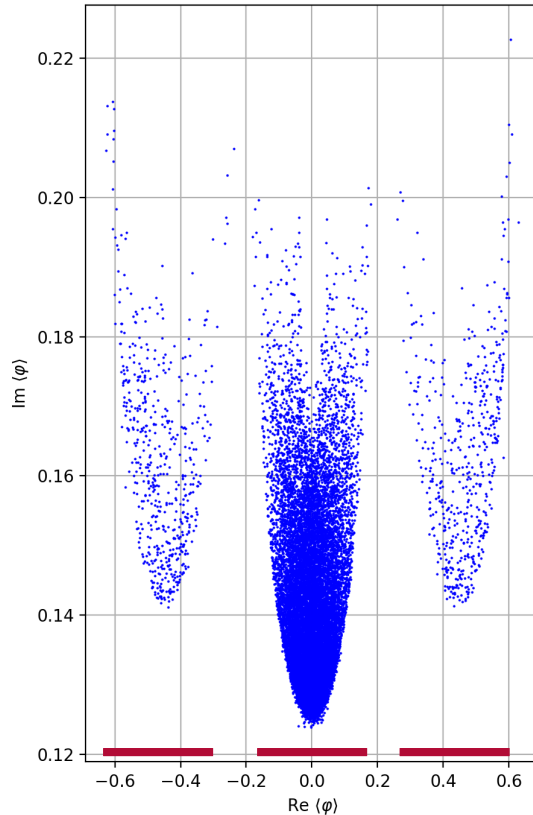
“activation function” f



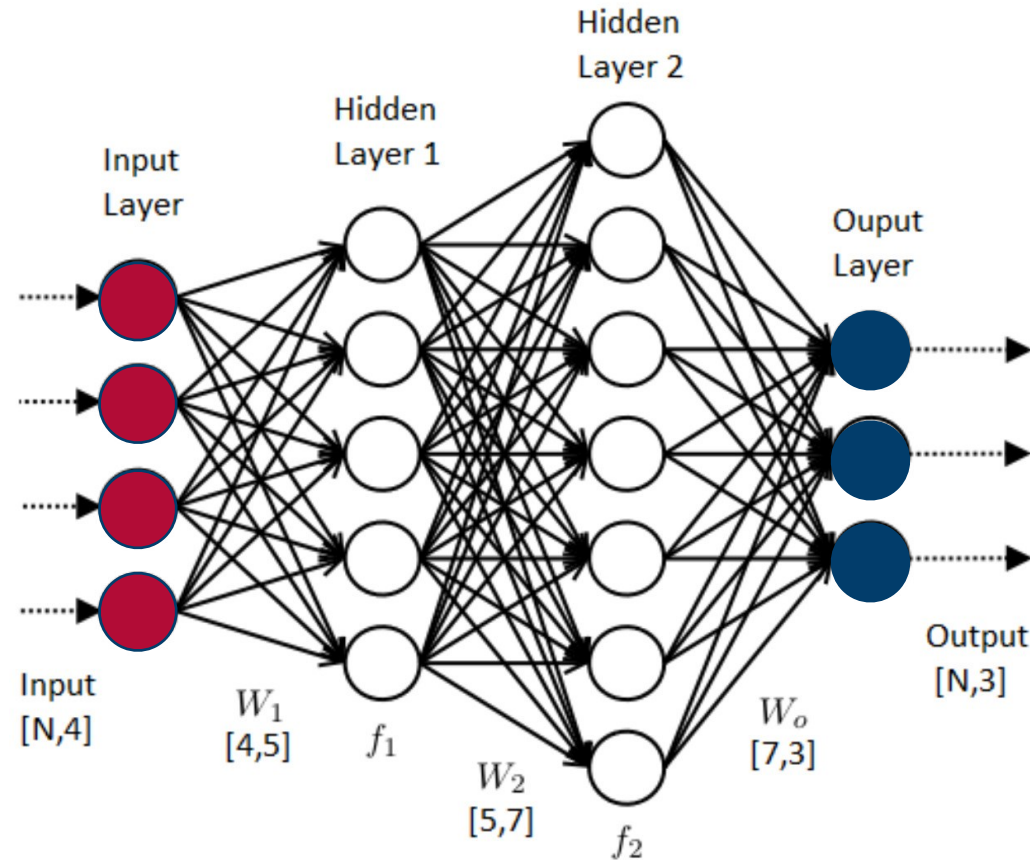
TRAINING THE NETWORK

Trying to get a computer to do what you want. . .

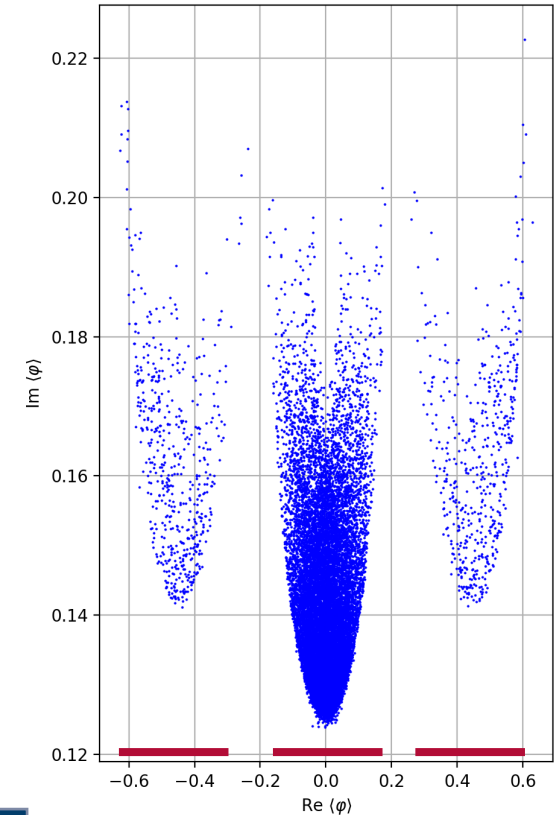
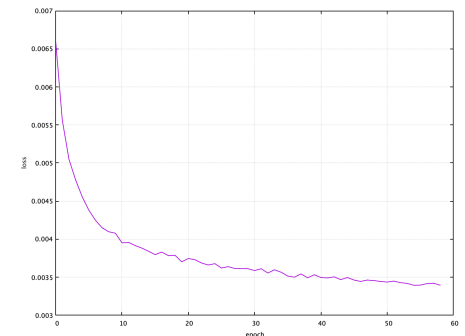
Single dense layer
“softmax” activation



training data

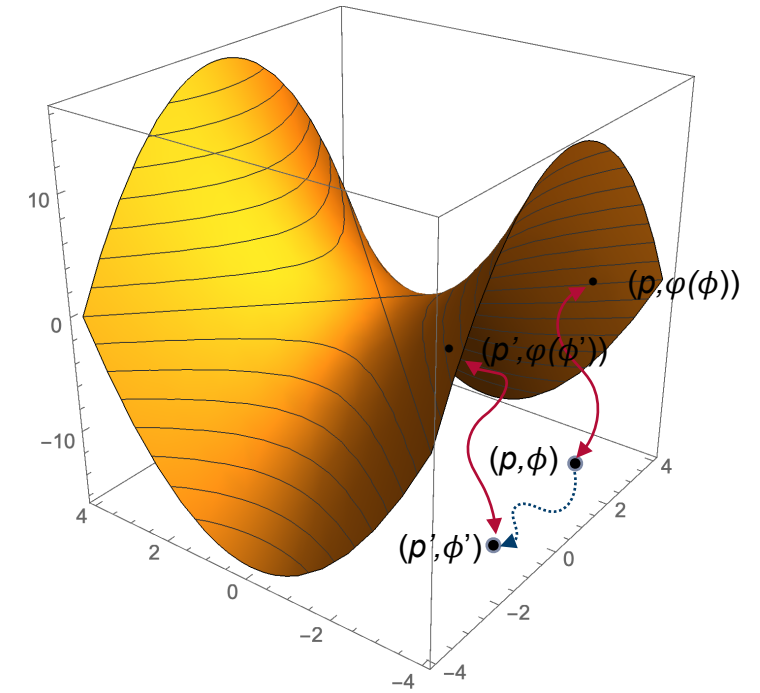


- Minimization of loss function
- Parameters tuned via stochastic gradient descent
- “Supervised training”

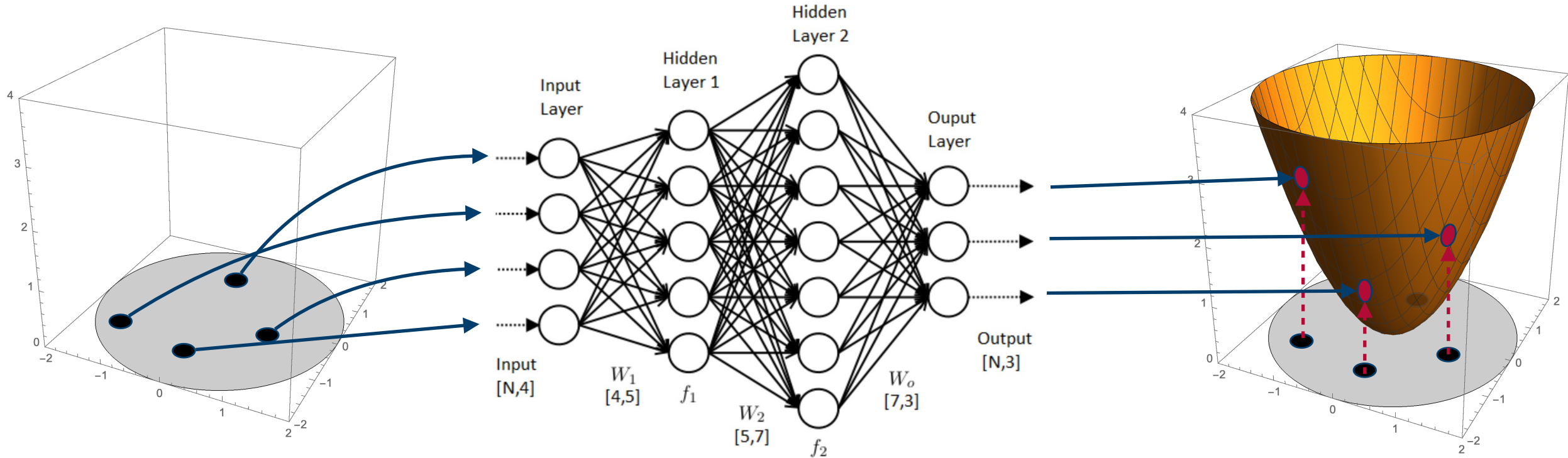


HMC WITH NEURAL NETWORKS

- Perform HMC on the real plane to make proposal $(p, \phi) \rightarrow (p', \phi')$
- $\phi' \rightarrow \text{NN} \rightarrow \varphi(\phi')$
- Accept/reject with $\mathbb{P}_{a/r} = \min \left(1, \frac{e^{-p'^2/2 - \text{Re}S_{eff}[\varphi(\phi')]} }{e^{-p^2/2 - \text{Re}S_{eff}[\varphi(\phi)]}} \right)$
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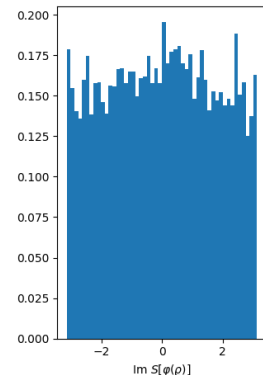
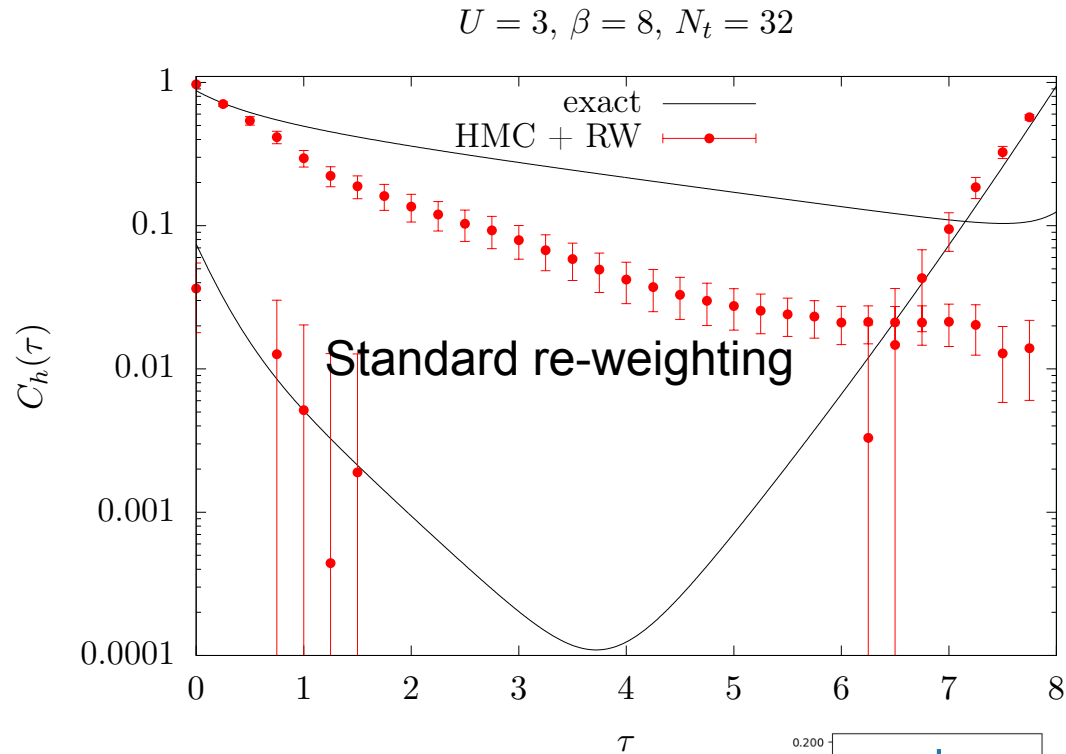


WHY IS THE NN USEFUL?

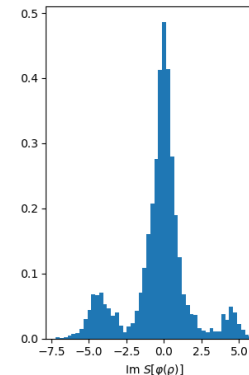
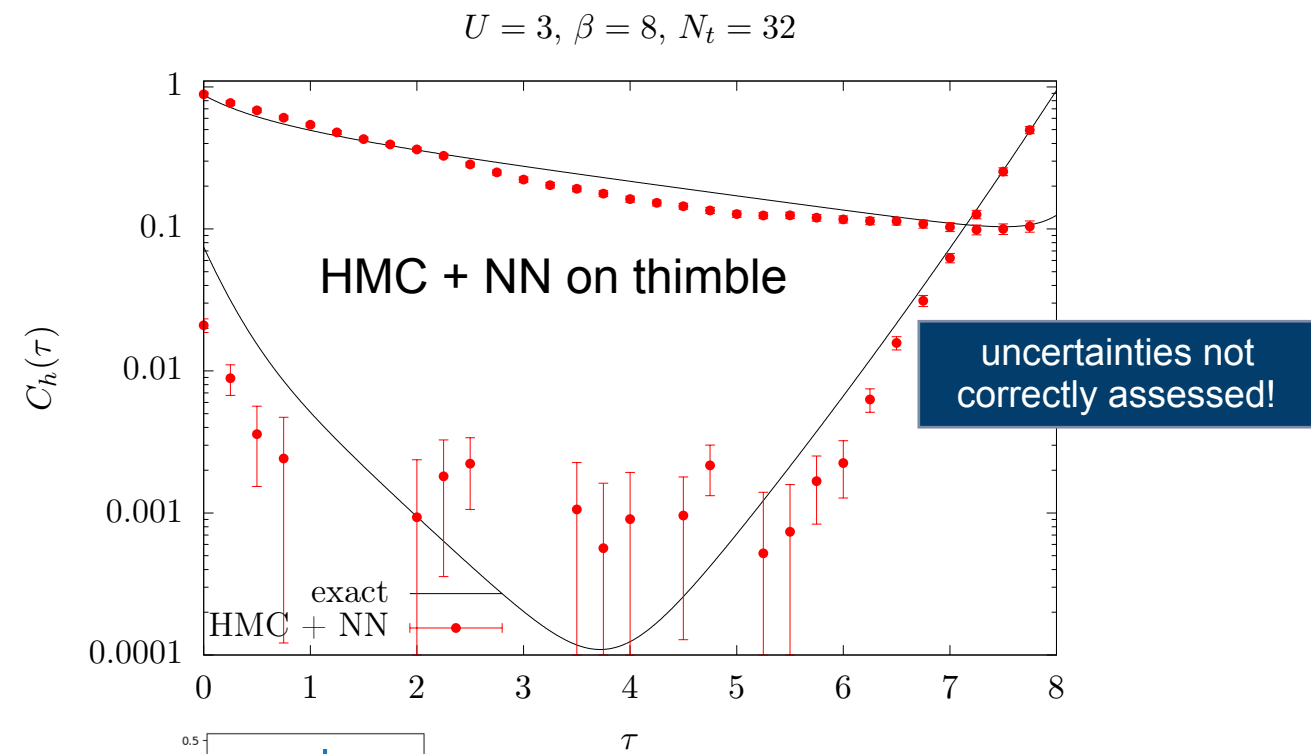


- Essentially a great big non-linear interpolation routine
 - Fast & efficient
- Calculation of Jacobian is *much* faster!
- Greatly alleviates ergodicity problem

SOME LIMITED SUCCESS SO FAR...

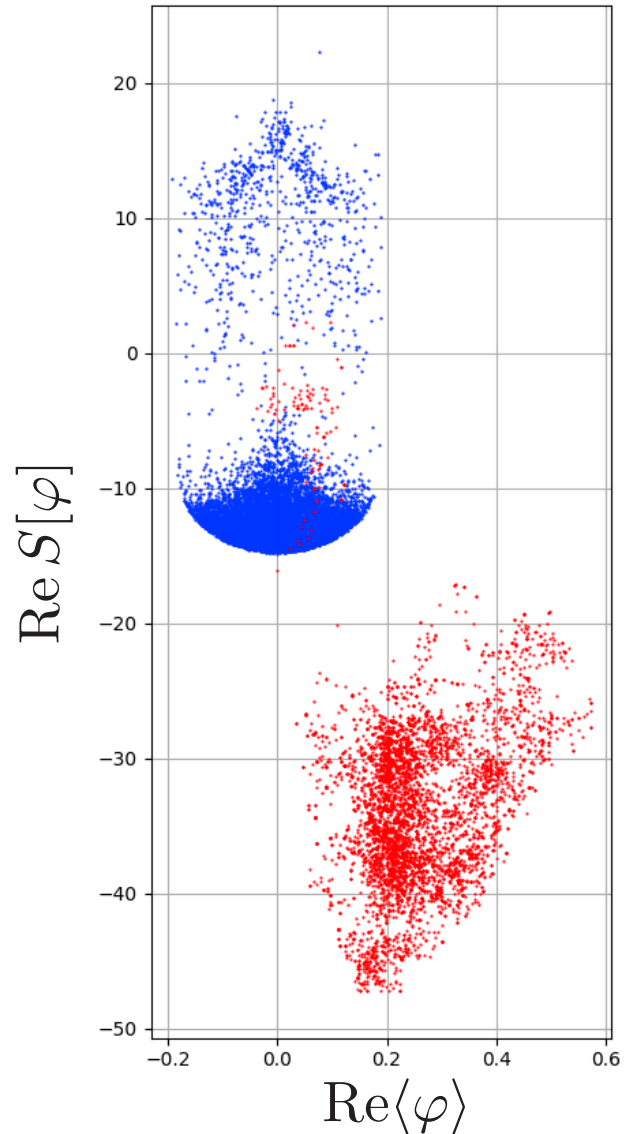


25

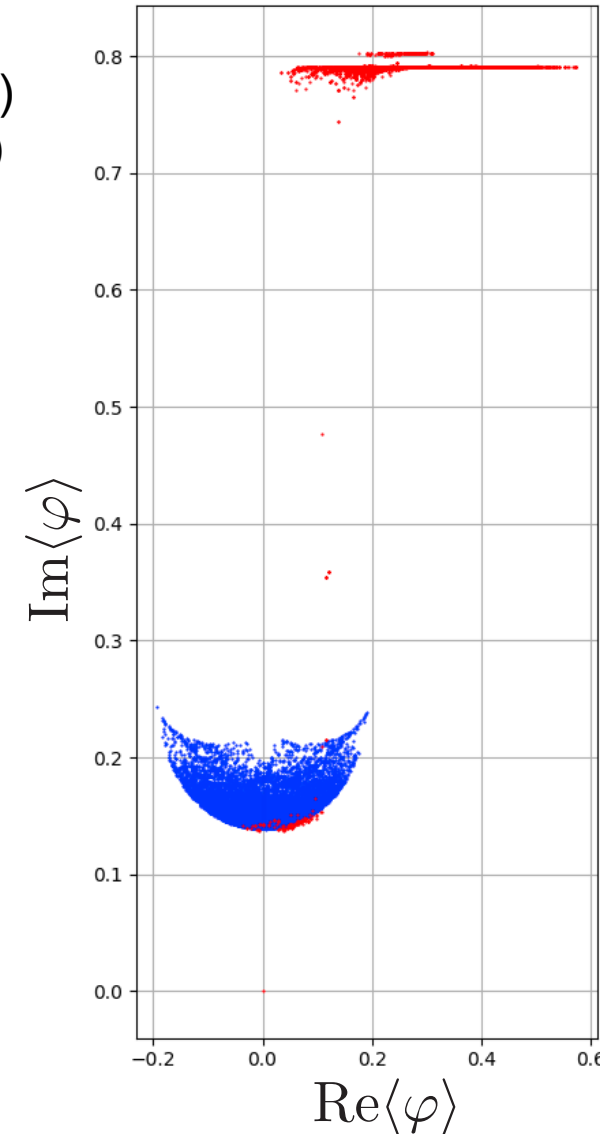


THE NN IS ONLY AS GOOD AS YOUR TRAINING DATA

Lessons learned so far. . .

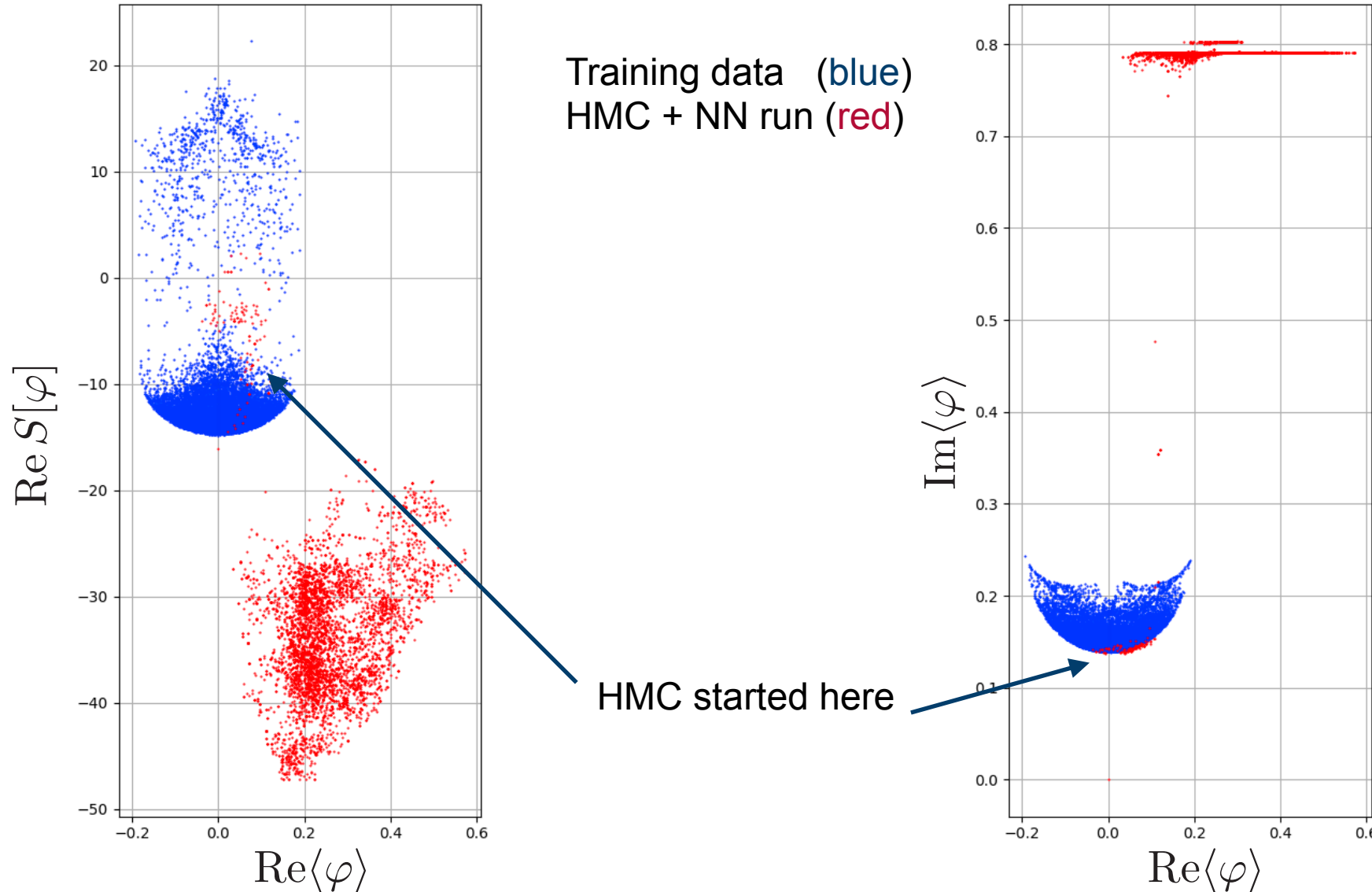


Training data (blue)
HMC + NN run (red)



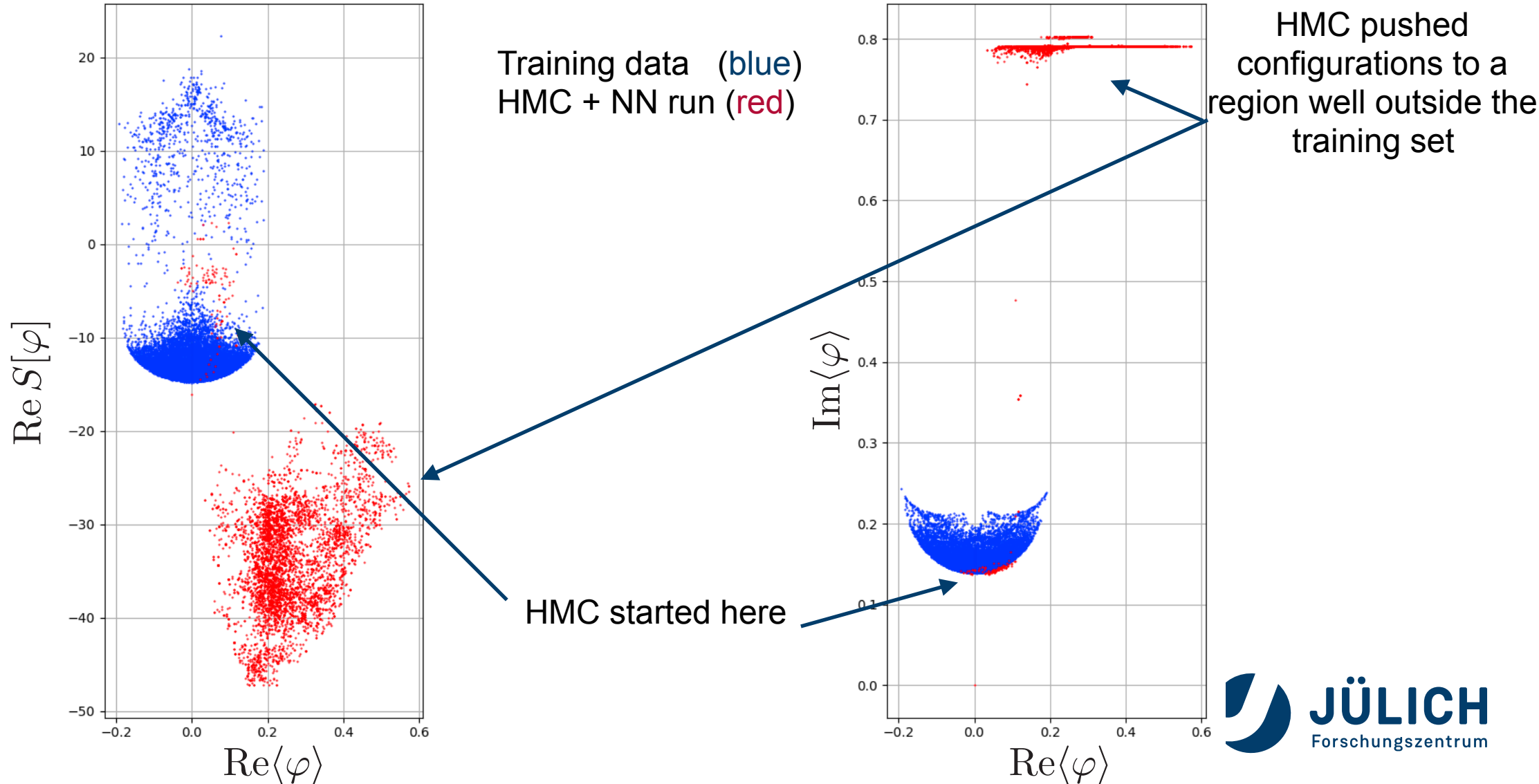
THE NN IS ONLY AS GOOD AS YOUR TRAINING DATA

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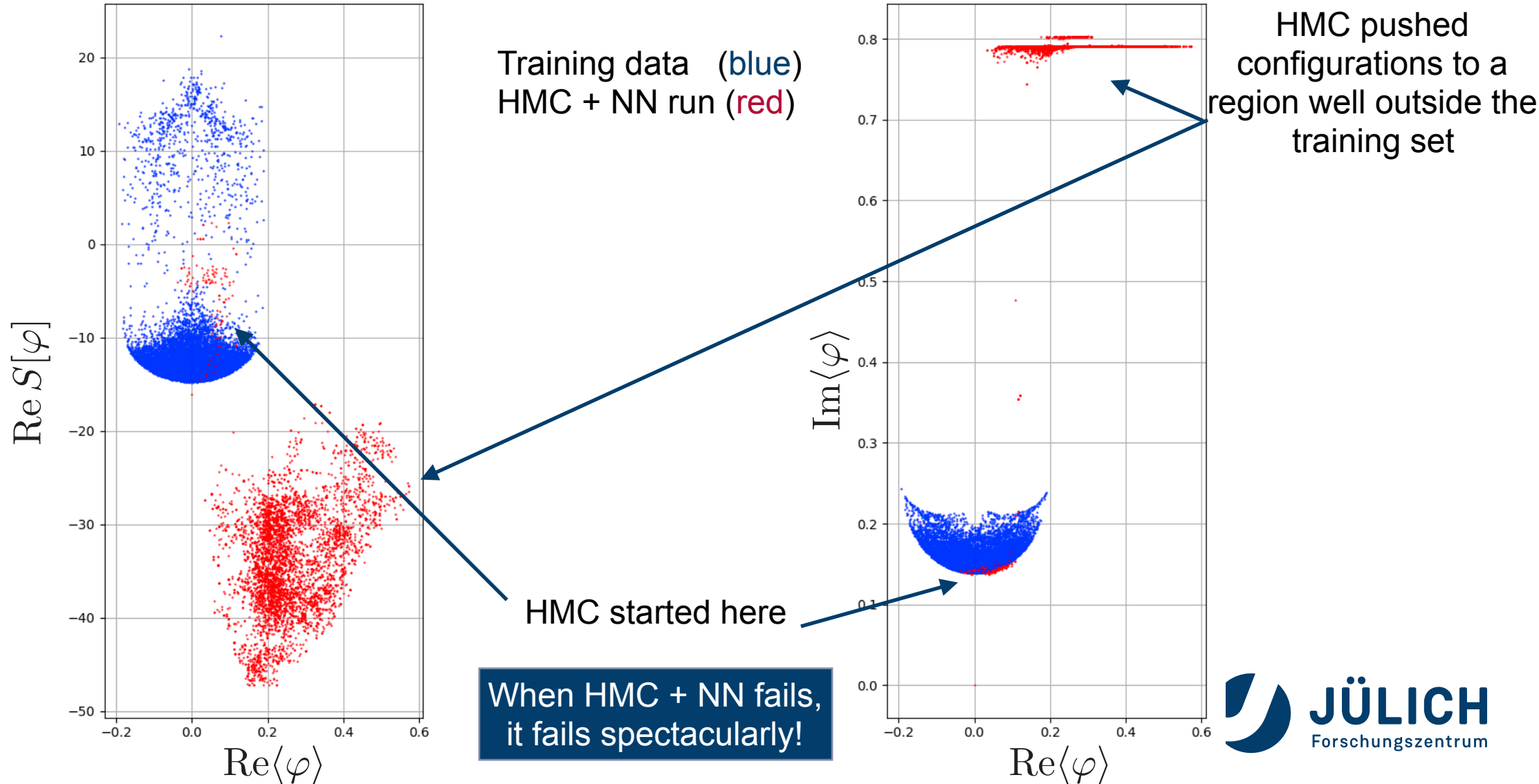
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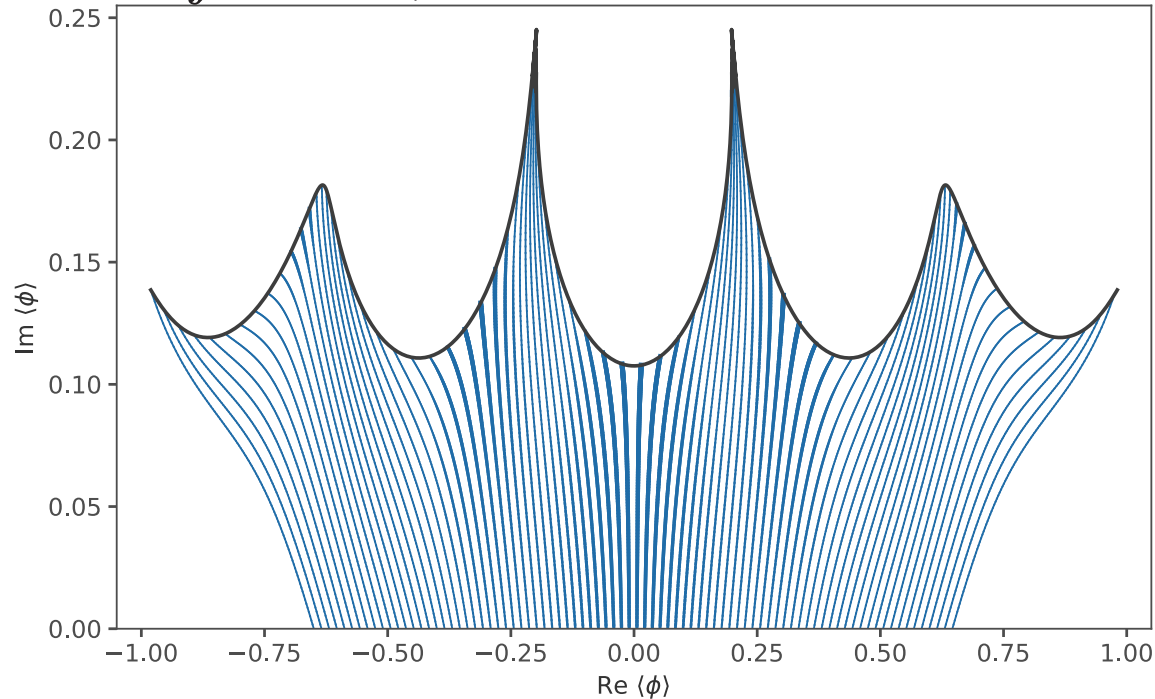
Lessons learned so far. . .



THE GOAL IS TO *ALLEVIATE* THE SIGN PROBLEM, NOT *SOLVE IT!*

Lessons learned so far. . .

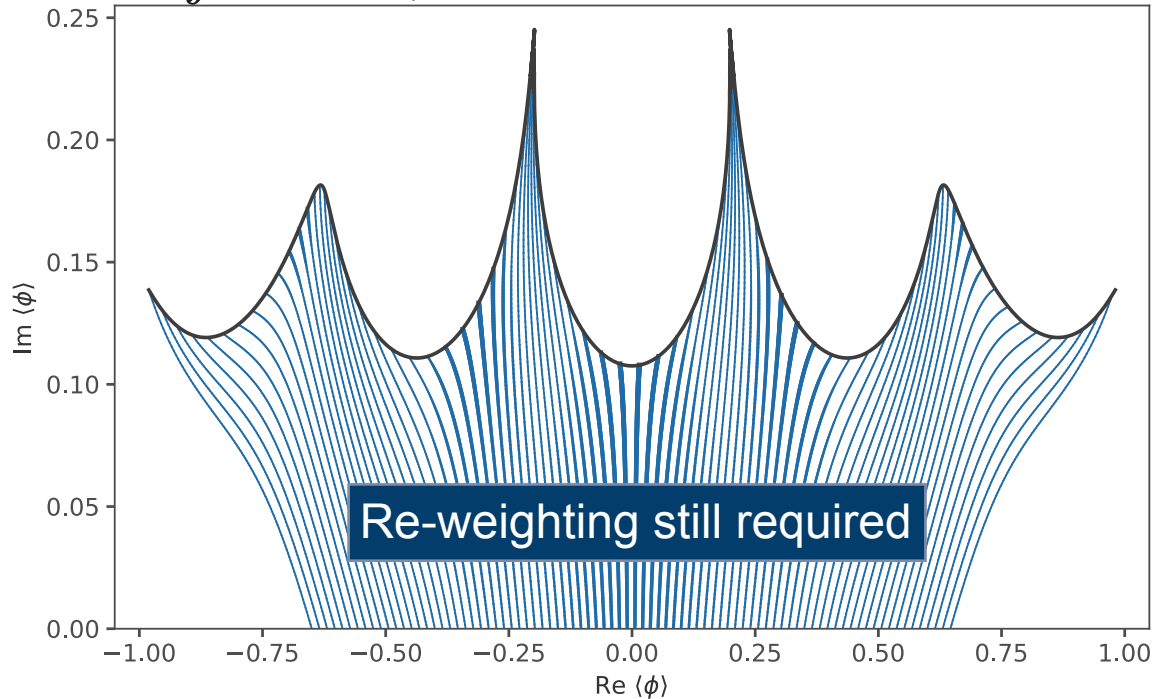
$$T_f = 1/2$$



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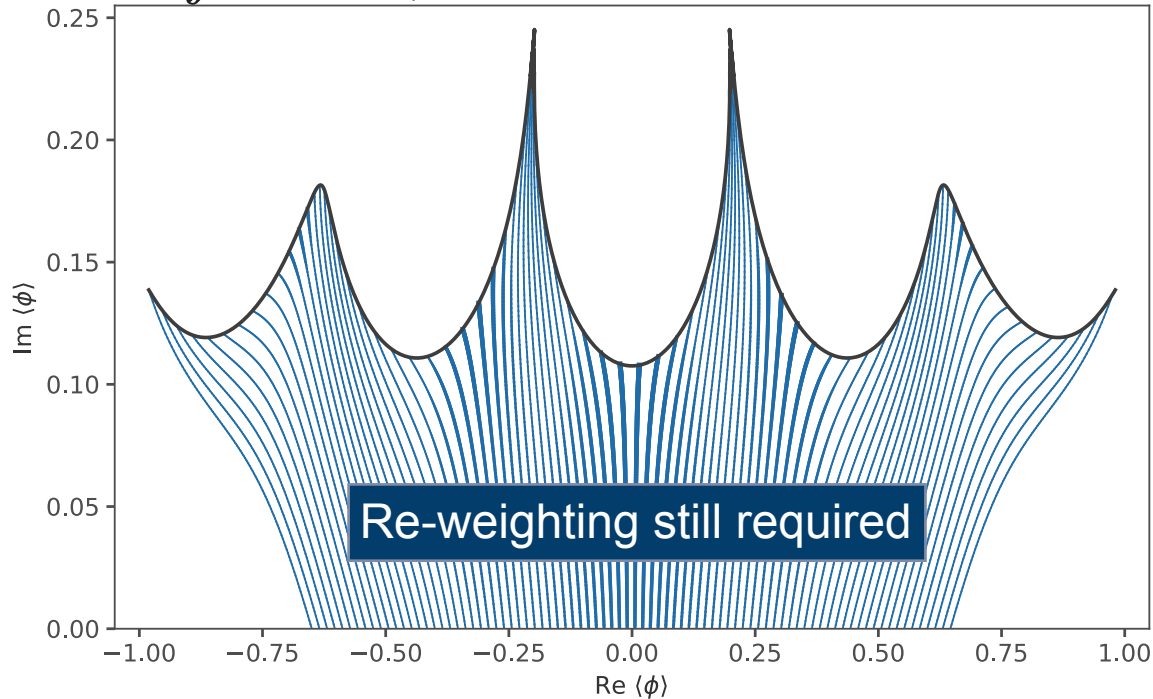
$$T_f = 1/2$$



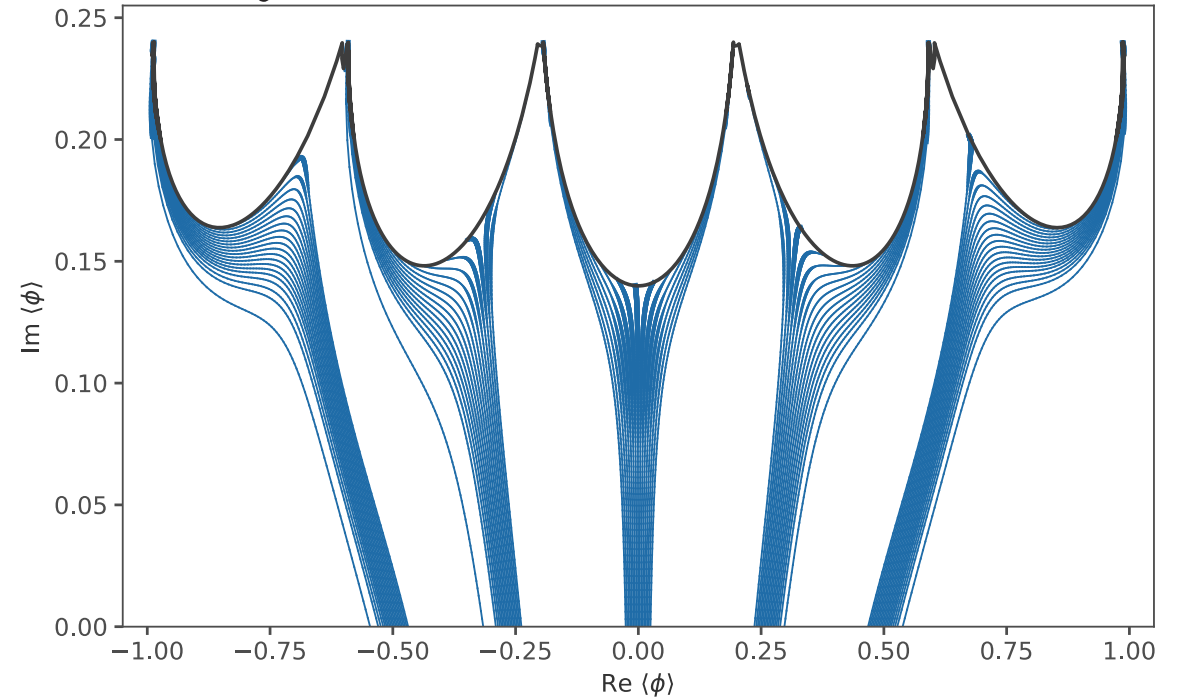
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Lessons learned so far. . .

$$T_f = 1/2$$



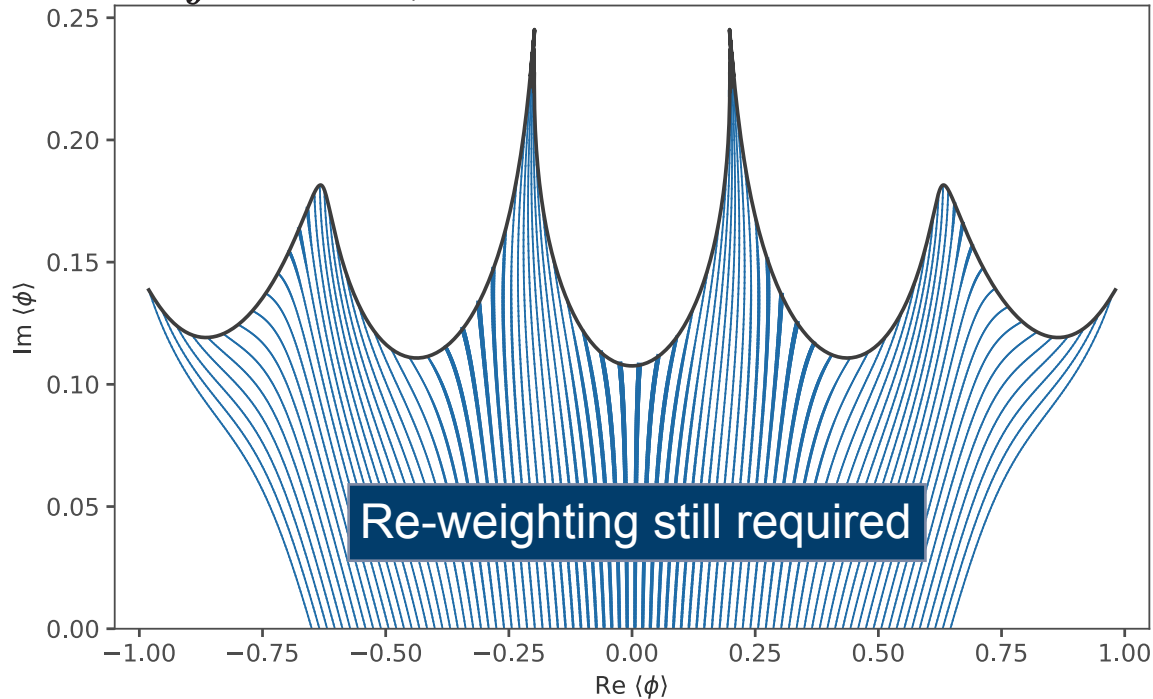
$$T_f = 1$$



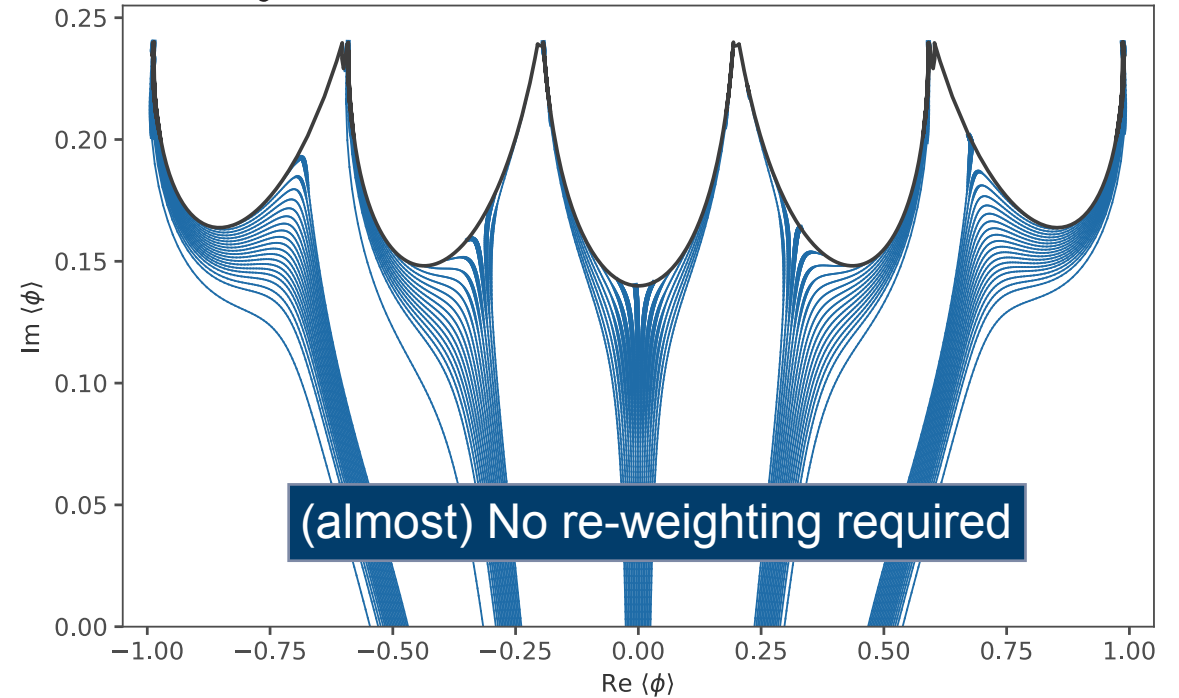
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Lessons learned so far. . .

$$T_f = 1/2$$



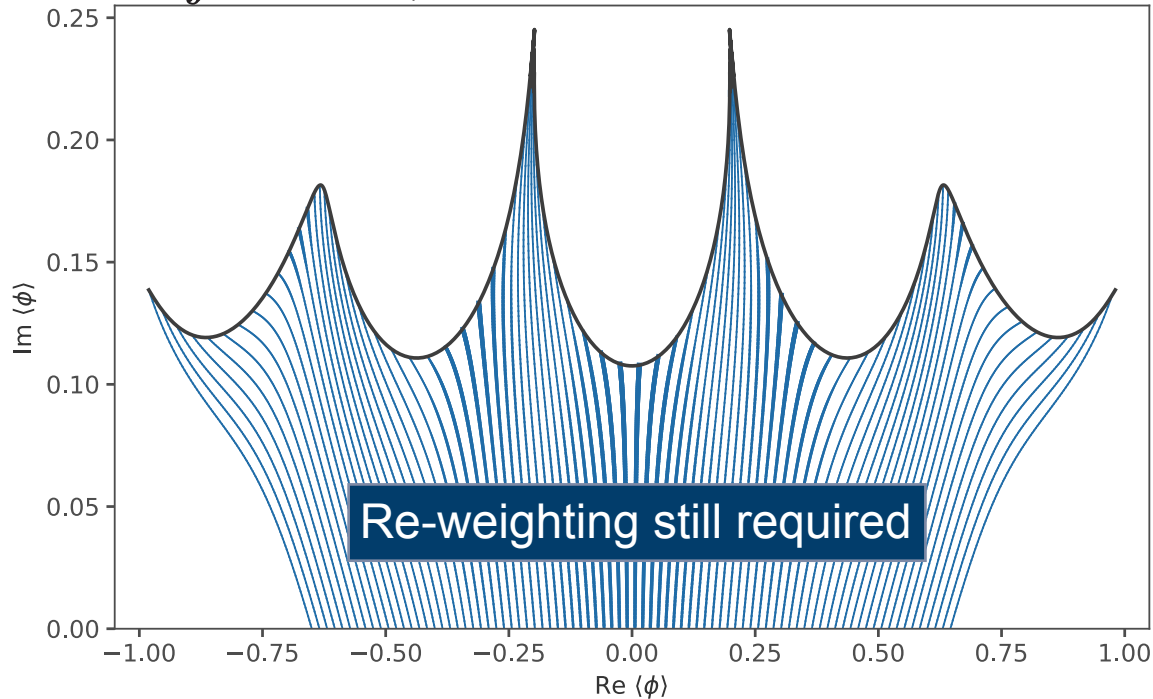
$$T_f = 1$$



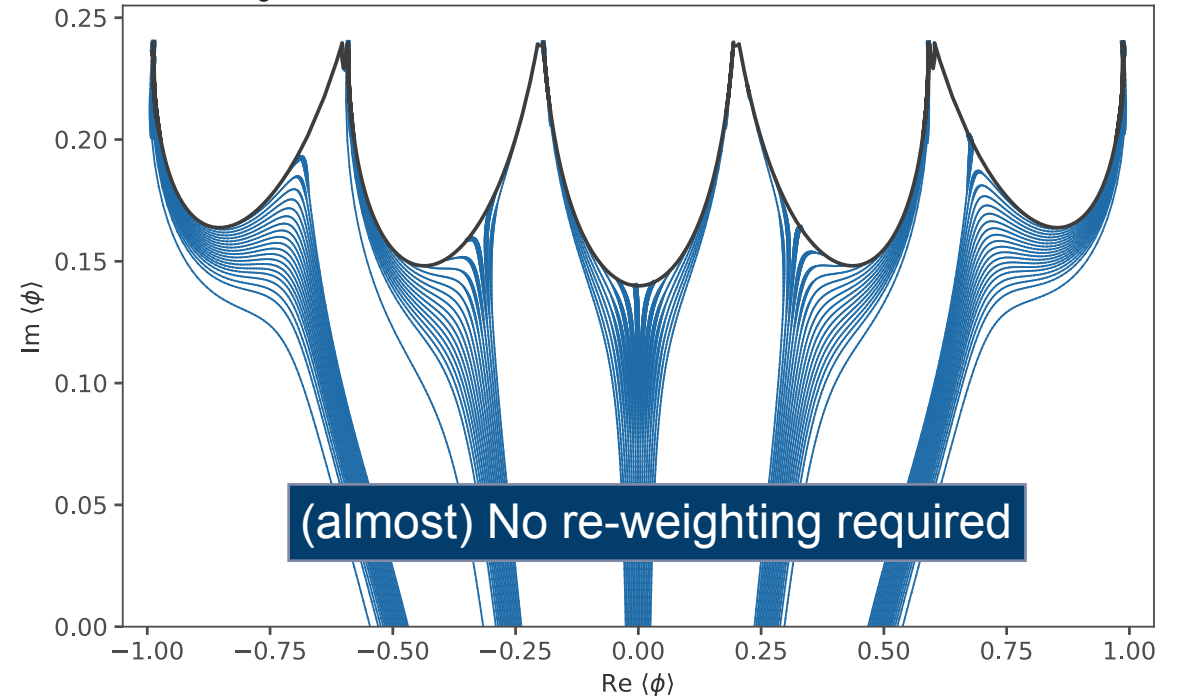
THE GOAL IS TO *ALLEVIATE* THE SIGN PROBLEM, NOT *SOLVE* IT!

Lessons learned so far...

$$T_f = 1/2$$



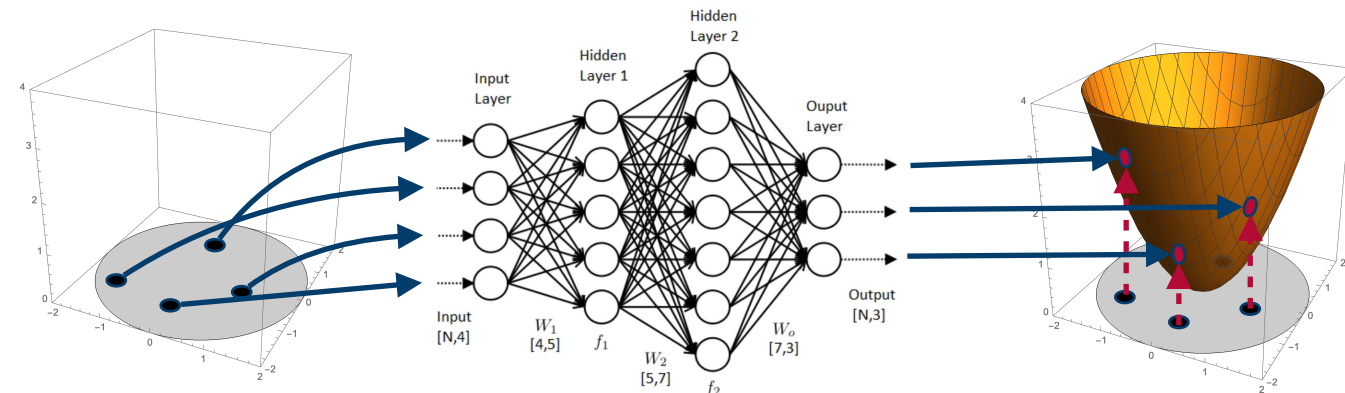
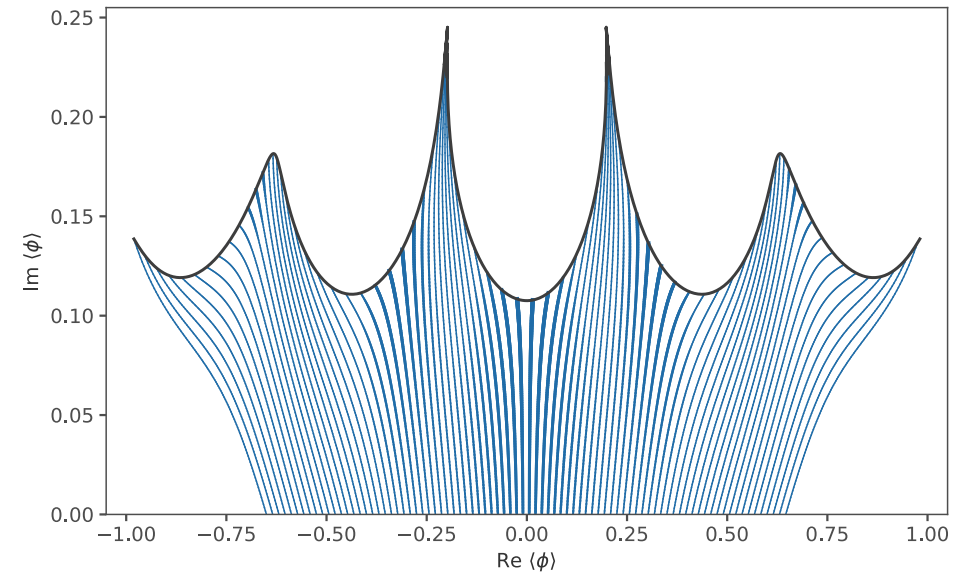
$$T_f = 1$$



But!!! Numerically very difficult for HMC+NN to stay on the thimble

CONCLUSIONS

- Sign problem can be alleviated
 - by holomorphic flow
 - calculation of Jacobian prohibitive (numerically)
- Used NN to train location of manifolds, i.e.. ML
 - Fast/efficient interpolation routine
 - Calculation of Jacobian much faster
- Found success in simple systems, but. . .
 - Still understanding limits of ML
 - Bias/limitation of training data
- Go to larger systems, like C20, C60, doped tubes
- Ultimate goal: apply holomorphic flow to lattice gauge theories



Thank you!

FLOW EQUATIONS

$$\phi_i = x_i + iy_i$$

$$\dot{\phi} = -(\nabla_{\phi} S[\phi])^* \implies$$

$$\frac{dx_i}{dt} =$$

$$\frac{dy_i}{dt} =$$

$$-\frac{\partial S_R}{\partial x_i}$$

$$-\frac{\partial S_R}{\partial y_i}$$

=

=

$$-\frac{\partial S_I}{\partial y_i},$$

$$\frac{\partial S_I}{\partial x_i}.$$

Gradient flow (heat equation)

EoM (constant of motion)

