

Constant path integral contour shifts

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Bern, January 23, 2025



Gefördert durch

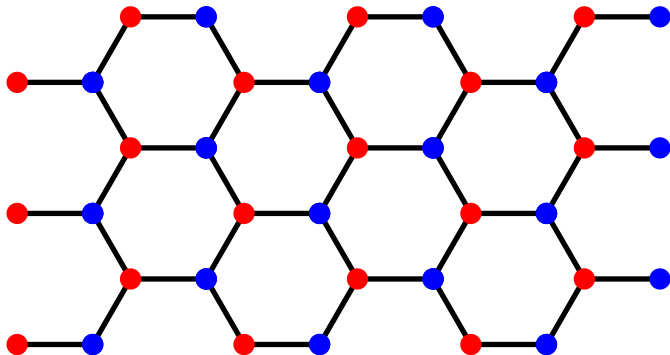


Deutsche
Forschungsgemeinschaft

Hubbard model

[Hubbard *ProcRSoc* **276** (1963); Novoselov + *Science* **306** (2004); Wallace *PhysRev* **71** (1947)]

$$H = - \sum_{\langle x,y \rangle, s} c_{x,s}^\dagger c_{y,s} + \frac{1}{2} U \sum_x q_x^2$$



Beyond half filling?

Sign problem!

$$H = - \sum_{\langle x,y \rangle, s} c_{x,s}^\dagger c_{y,s} + \frac{1}{2} U \sum_x q_x^2 + \mu \sum_x q_x$$

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$$p[\phi] \propto \det (M[\phi, \mu] M[\phi, -\mu]^\dagger) \not\geq 0$$

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► Lefschetz thimbles & contour deformation

[Alexandru + *PRD* **93** (2016); Cristoforetti + *PRD* **88** (2013); Rodekamp, JO + *PRB* **106** (2022);

Ulybyshev + *PRD* **101** (2020); Wynen, JO + *PRB* **103** (2021)]

[Gäntgen, JO + *PRB* **109** (2024); Rodekamp, JO + *EPJB* accepted (2024)]

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► Tensor Networks

[Corboz *PRB* **93** (2016); Schneider, JO + *PRB* **104** (2021)]

[Suladze, JO + (forthcoming)]

The Team



Evan
Berkowitz



Stefan Krieg



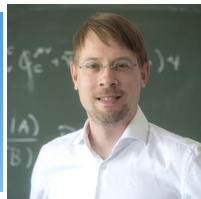
Thomas Luu



Giovanni
Pederiva



Manuel
Schneider



Carsten
Urbach



[Gäntgen, JO +
PRB 109 (2024)]

Christoph Gäntgen



[Rodekamp, JO +
EPJB accepted
(2024)]

Marcel Rodekamp



[Suladze, JO +
(forthcoming)]

Archil Suladze

Disclaimers

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1. Typically not the best, but versatile approach.

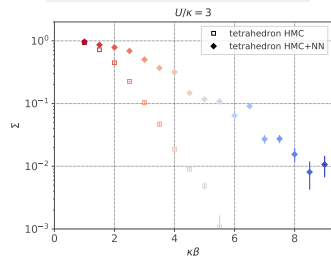
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Disclaimers

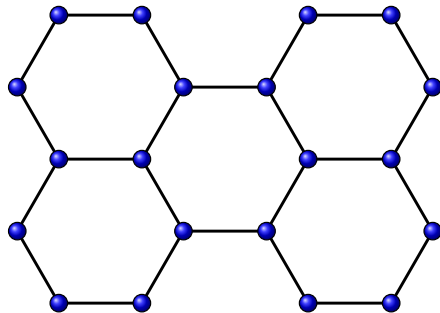
1. Typically not the best, but versatile approach.
2. Sign problem still exponentially bad.



[Wynen, JO + *PRB* 103 (2021)]

Perylene

[Cao & Yang *RSC Adv* **12** (2022); Sato + *IEEE JSelTopQEI* **4** (1998); Shchuka + *ChemPhysLet* **164** (1989)]



Candidate for organic LEDs, solar cells,...

Lefschetz thimbles

[Alexandru + *PRD* **93** (2016); Lefschetz *AMS* **22** (1921); Tanizaki + *NewJPhys* **18** (2016); talk by M. Ulybyshev]

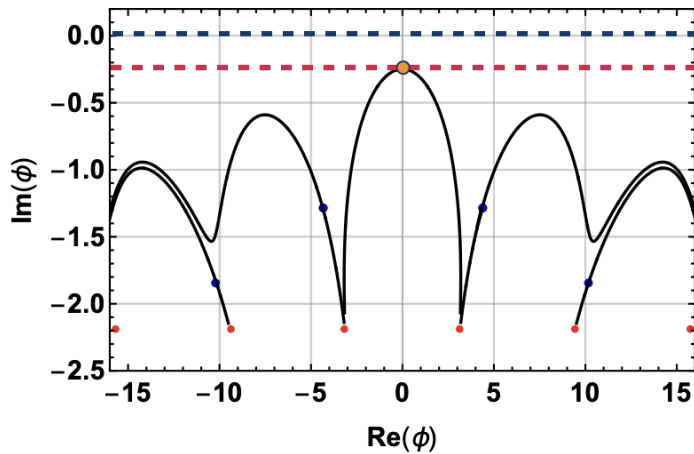
► Path integral formalism

$$\phi \sim e^{-S[\phi]}$$

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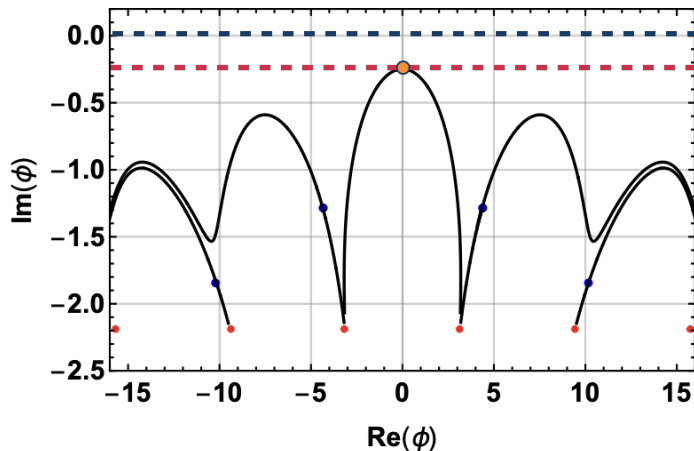
- ▶ Path integral formalism
 $\phi \sim e^{-S[\phi]}$
- ▶ Complex manifolds of constant $\text{Im}S$



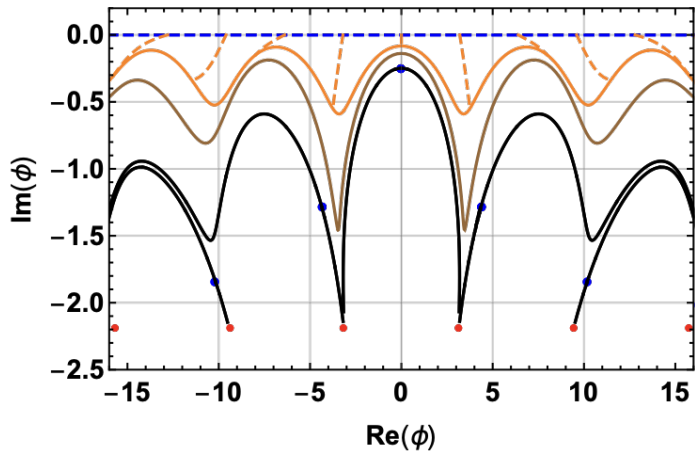
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- ▶ Path integral formalism
 $\phi \sim e^{-S[\phi]}$
- ▶ Complex manifolds of constant $\text{Im}S$
- ▶ Same path integral by Cauchy's theorem

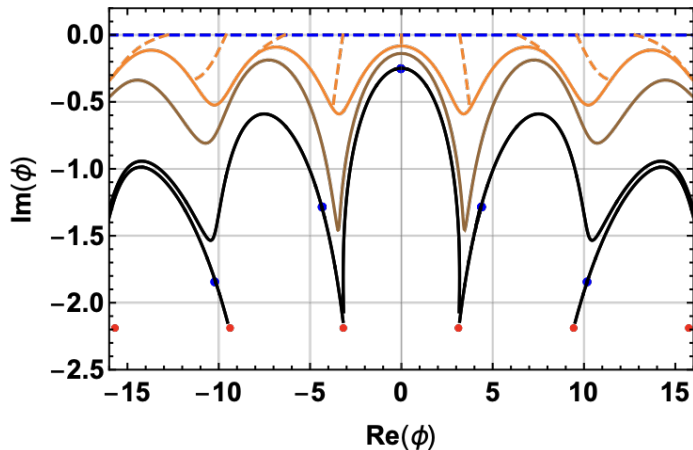


Holomorphic flow [Cristoforetti + PRD 86 (2012)]



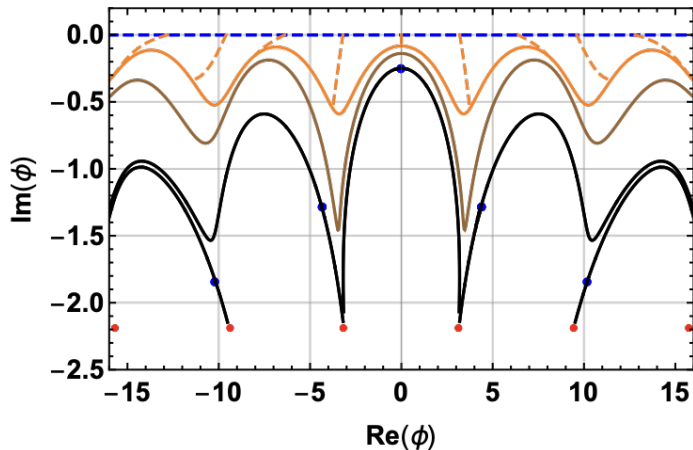
Holomorphic flow [Cristoforetti + PRD 86 (2012)]

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Extremely expensive!

Use Machine Learning

[Alexandru + *PRD* **96** (2017); Rodekamp, JO + *PRB* **106** (2022); Wynen, JO + *PRB* **103** (2021)]

- ▶ Flow some random field configurations

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- ▶ Flow some random field configurations
- ▶ ‘Learn’ structure of Lefschetz Thimbles from flowed data

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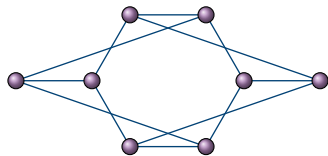
- ▶ Flow some random field configurations
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- ▶ SHIFT: $\mathbb{R}^n \rightarrow \mathbb{C}^n, \phi \mapsto \mathcal{NN}(\phi)$

Use Machine Learning

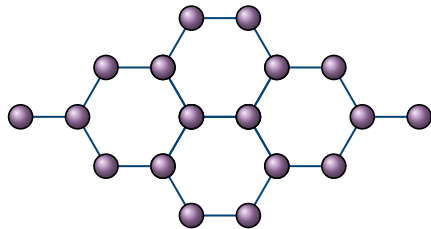
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- ▶ ‘Learn’ structure of Lefschetz Thimbles from flowed data
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- ▶ Apply reweighting \Rightarrow SHIFT doesn't have to be perfect

Benchmark lattices [Rodekamp, JO + PRB 106 (2022)]

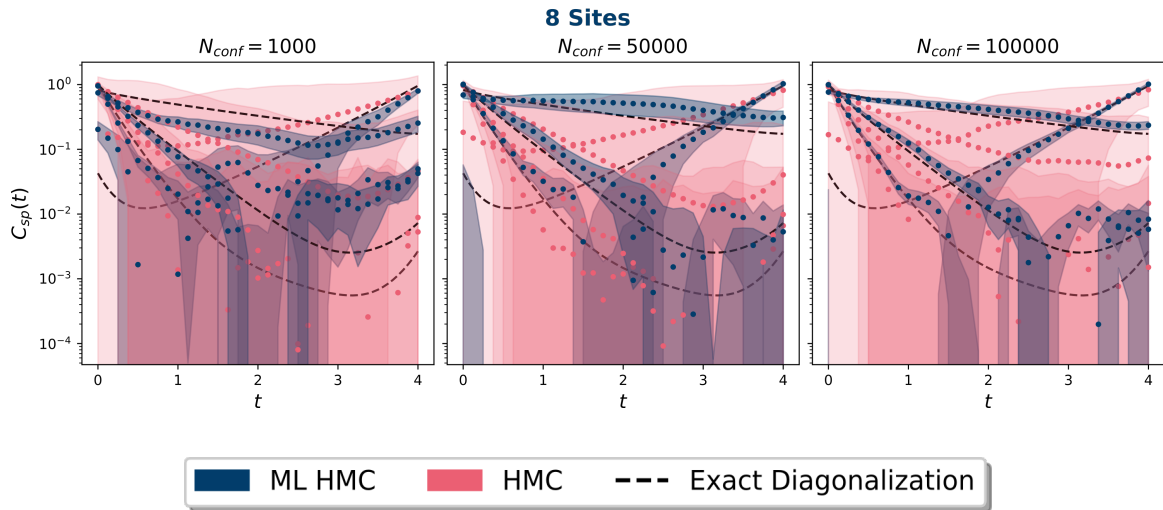


(a) 8 Sites

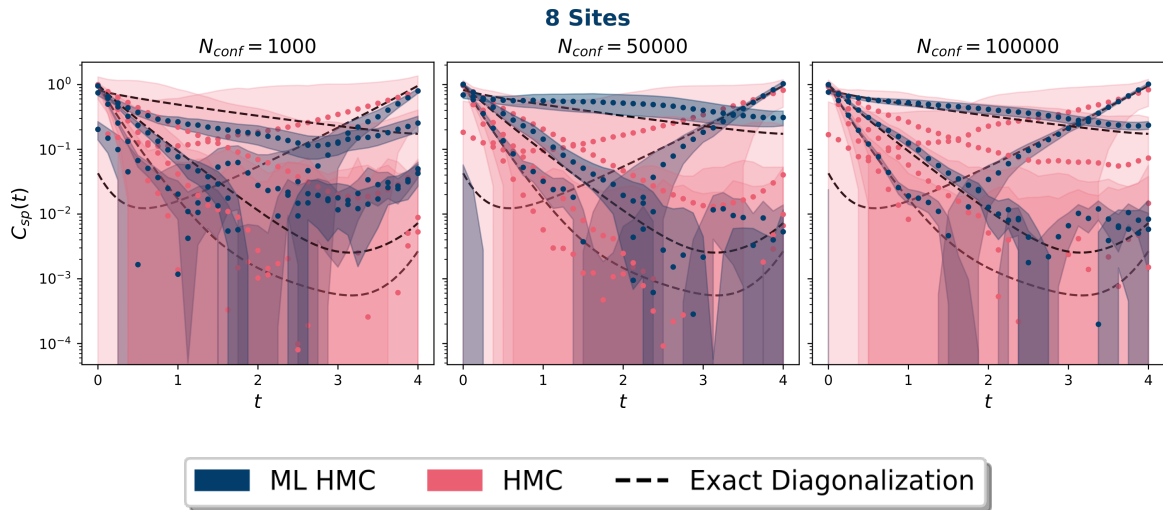


(b) 18 Sites (boundary suppressed)

Single Particle Correlators [Rodekamp, JO + PRB 106 (2022)]

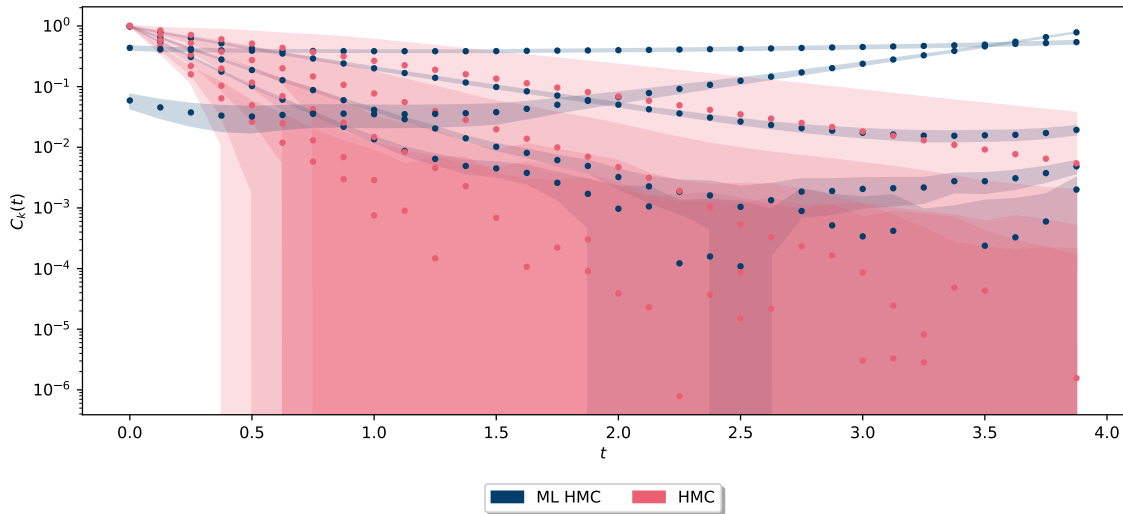


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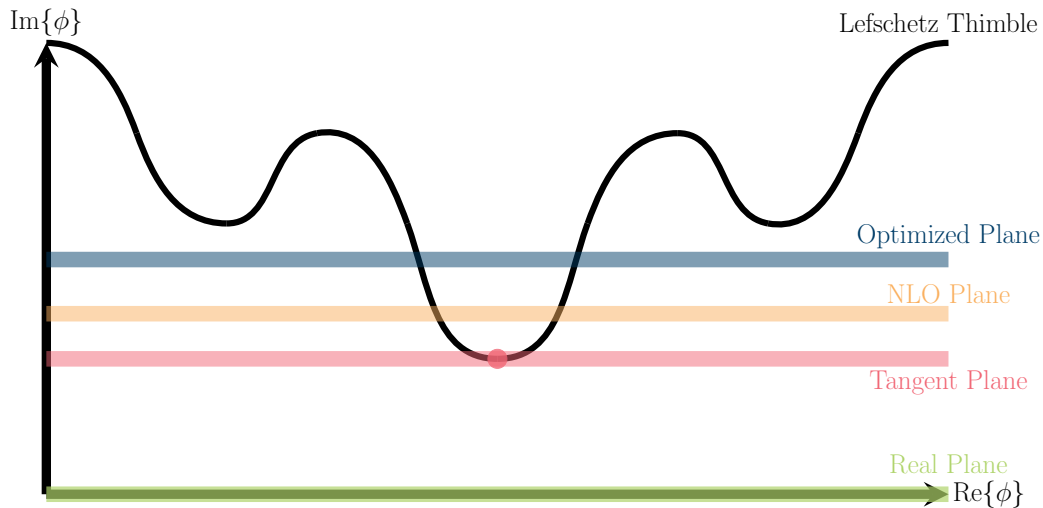


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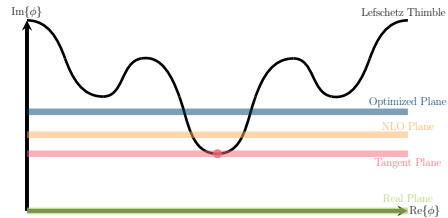
18 Sites



Sketch of manifolds [Gäntgen, JO + PRB 109 (2024)]

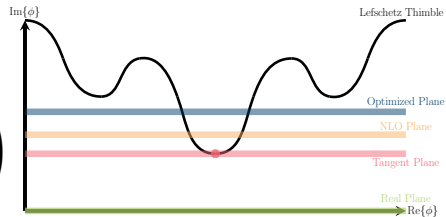


Analytic formulae [Gäntgen, JO + PRB 109 (2024)]



Tangent plane (saddle point of S):

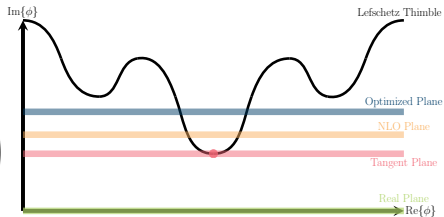
$$\phi_0/\delta = -\frac{U}{N_x} \sum_k \tanh \left(\frac{\beta}{2} [\epsilon_k + \mu + \phi_0/\delta] \right)$$



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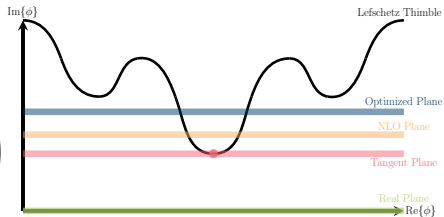
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Next-to-leading order (NLO) plane
(saddle point of $S_{\text{eff}} = S + \frac{1}{2} \log \det \mathbb{H}$)

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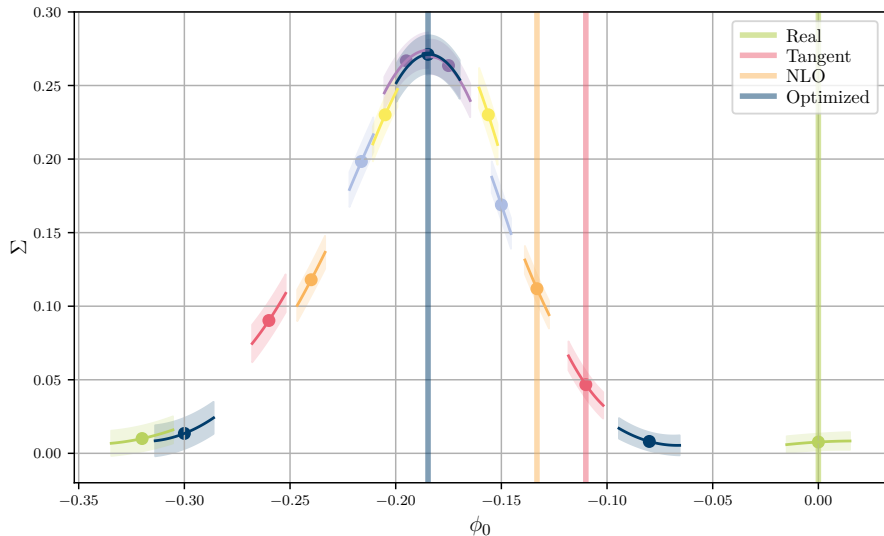


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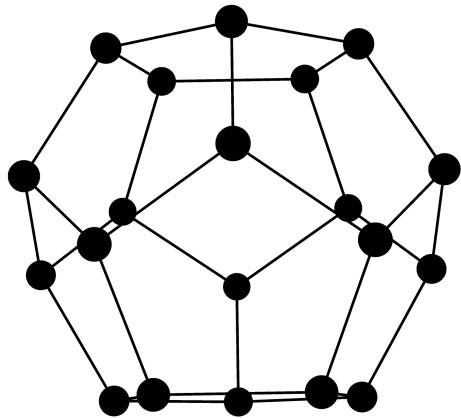
$$\mathbb{H}_{x't',xt} = \left(\frac{1}{\tilde{U}} - 1 \right) \delta_{x',x} \delta_{t',t} - T_{+;xt,x't'} T_{+;x't',xt} - T_{-;xt,x't'} T_{-;x't',xt} ,$$

$$T_{\pm;x't',xt} = \sum_{kn} \Lambda_{x't',kn}^\dagger \frac{e^{\pm(\delta\epsilon_k + \delta\mu + \phi_1 + i\tilde{\omega}_n)}}{1 - e^{\pm(\delta\epsilon_k + \delta\mu + \phi_1 + i\tilde{\omega}_n)}} \Lambda_{kn,xt}$$

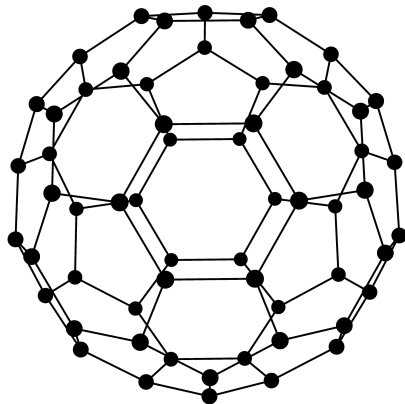
Numerical search [Gäntgen, JO + PRB 109 (2024)]



Bigger benchmark lattices [Gäntgen, JO + PRB 109 (2024)]

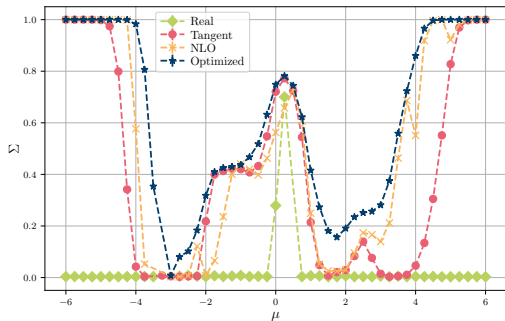


C_{20}

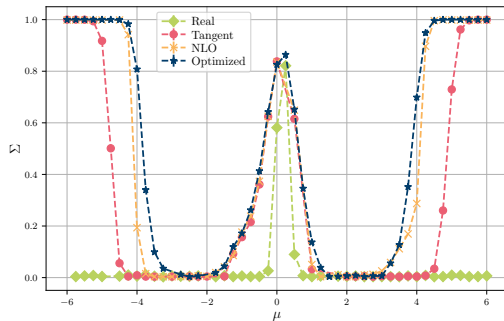


C_{60}

Statistical power [Gäntgen, JO + PRB 109 (2024)]



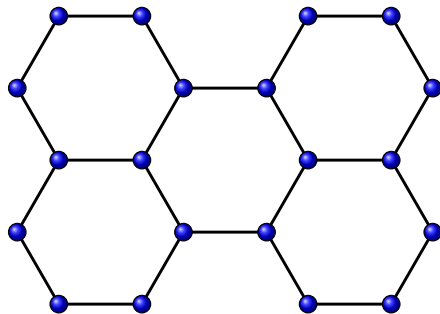
C_{20}



C_{60}

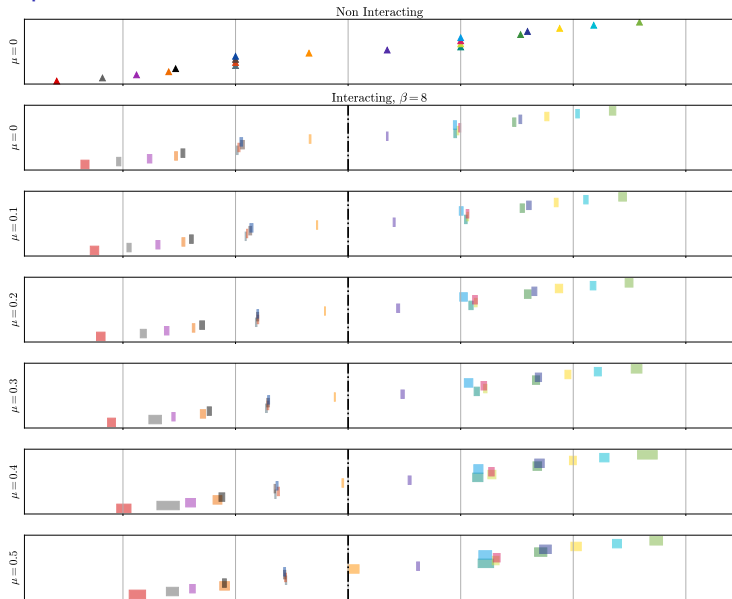
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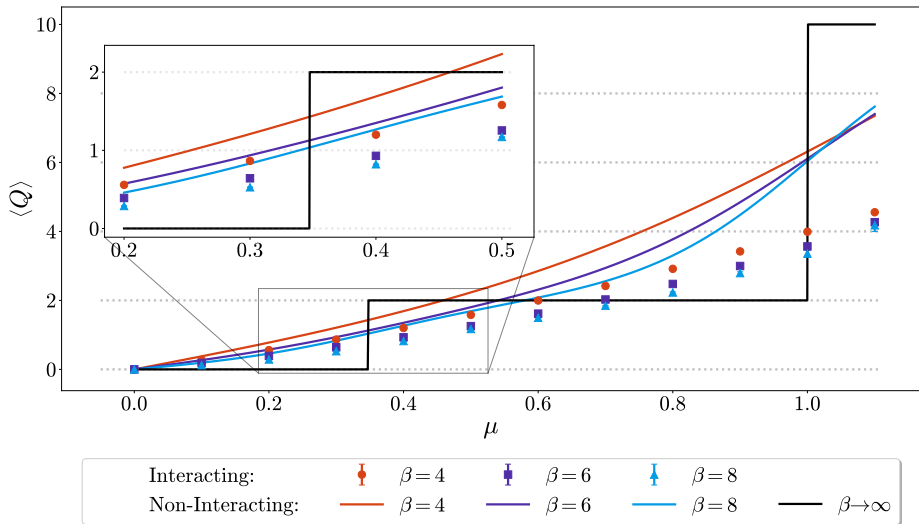


Candidate for organic LEDs, solar cells,...

Single particle spectrum [Rodekamp, JO + EPJB accepted (2024)]



Charge with doping [Rodekamp, JO + EPJB accepted (2024)]



+ SIGN25 +

Tensor Networks (TN) have no sign problem [Chen + *PRXQuantum* **6** (2025)]

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Right!

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Wrong! (kind off)

Contraction complexity depends on average sign.

$$\langle \dots \rangle = \text{tr} \left(\begin{array}{cccc} \dots & & & \\ & \dots & & \\ & & \dots & \\ & & & \dots \end{array} \right)$$

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Projected Entangled Pair States (PEPS)

[Orús *AnnPhys* **349** (2014); Verstraete & Cirac *cond-mat/0407066*]

$$|\psi\rangle = \sum_{s_1} \sum_{s_2} \cdots \sum_{s_N} A_{s_1, s_2, \dots, s_N} |s_1\rangle \otimes |s_2\rangle \otimes \cdots \otimes |s_N\rangle$$

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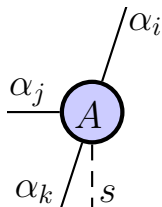
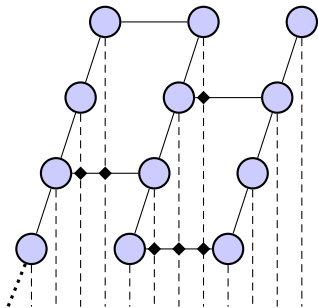
$$\begin{aligned} |\psi\rangle &= \sum_{s_1} \sum_{s_2} \cdots \sum_{s_N} A_{s_1, s_2, \dots, s_N} |s_1\rangle \otimes |s_2\rangle \otimes \cdots \otimes |s_N\rangle \\ &\approx \sum_{s_1} \sum_{s_2} \cdots \sum_{s_N} A_{s_1; \alpha_1}^1 A_{s_2; \alpha_1, \alpha_2}^2 \cdots A_{s_N; \alpha_{N-1}}^N |s_1\rangle \otimes |s_2\rangle \otimes \cdots \otimes |s_N\rangle \end{aligned}$$

Truncate $\alpha_i \leq D \forall i$

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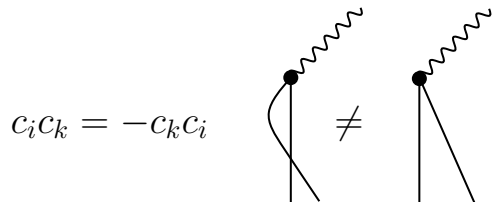
Truncate $\alpha_i \leq D \forall i$

Contractions [Schuch + PRL 98 (2007)]

$$d = 1 \quad \text{---} \bullet \text{---} \bullet \text{---} = \text{---} \bullet \text{---}$$

$$d > 1 \quad \begin{array}{c} \diagup \\ \diagdown \end{array} \bullet \text{---} \bullet \begin{array}{c} \diagup \\ \diagdown \end{array} = \begin{array}{c} \diagup \\ \diagdown \end{array} \bullet \begin{array}{c} \diagdown \\ \diagup \end{array}$$

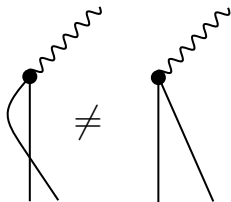
Fermionic PEPS [Corboz + *PRB* 81 (2010)]



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$$c_i c_k = -c_k c_i$$

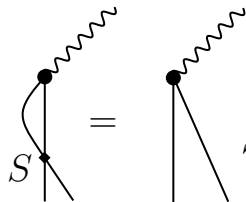
$$(c_i c_j) c_k = c_k (c_i c_j)$$



Fermionic PEPS [Corboz + PRB 81 (2010)]

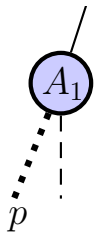
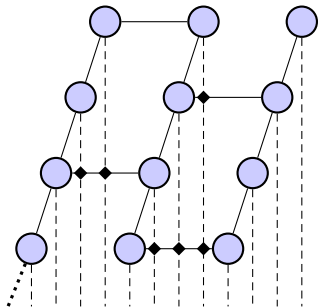
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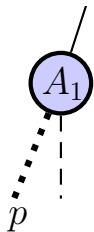
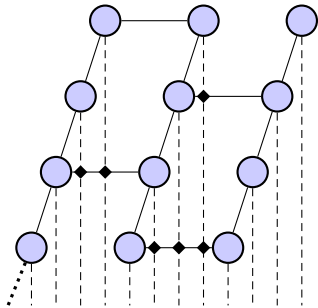


$$S = \begin{pmatrix} \overbrace{1 \dots 1}^{\text{even}} & \overbrace{1 \dots 1}^{\text{odd}} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 1 \dots 1 & 1 \dots 1 \\ 1 \dots 1 & -1 \dots -1 \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 1 \dots 1 & -1 \dots -1 \end{pmatrix} \left. \begin{array}{l} \text{even} \\ \text{odd} \end{array} \right\}$$

Parity link



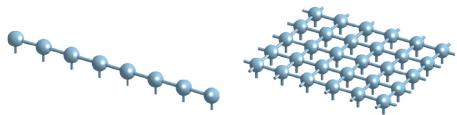
Parity link



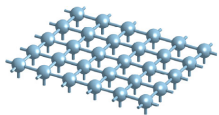
$$p = \pm 1$$

\Rightarrow even- and odd-parity
subspaces are disjoint

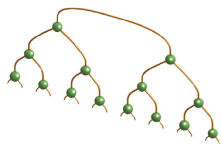
Augmented Tree Tensor Networks (aTTN) [Felser + *PRL* 126 (2021)]



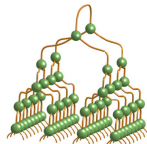
(a)



(b)



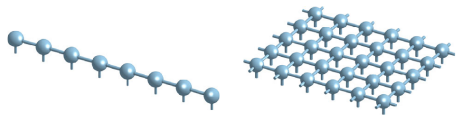
(c)



(d)

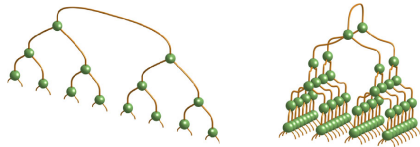
[Felser + *PRX* 10 (2020)]

Augmented Tree Tensor Networks (aTTN) [Felser + *PRL* 126 (2021)]



(a)

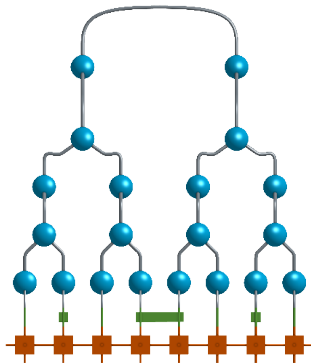
(b)



(c)

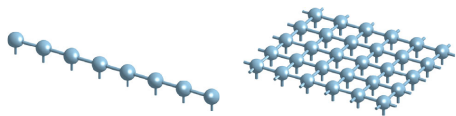
(d)

[Felser + *PRX* 10 (2020)]

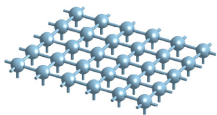


[Felser *PhD thesis* (2021)]

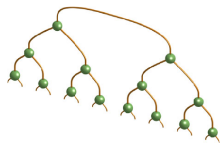
Augmented Tree Tensor Networks (aTTN) [Felser + *PRL* 126 (2021)]



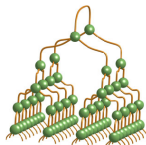
(a)



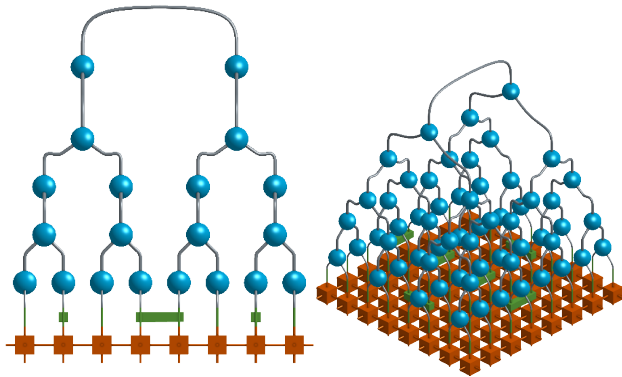
(b)



(c)



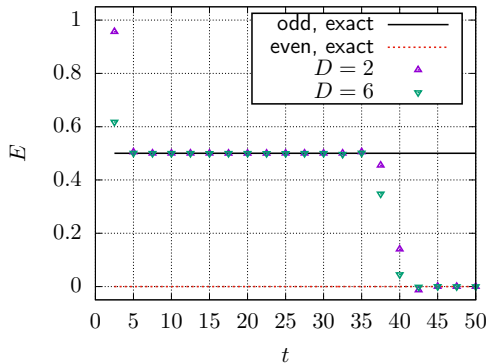
(d)



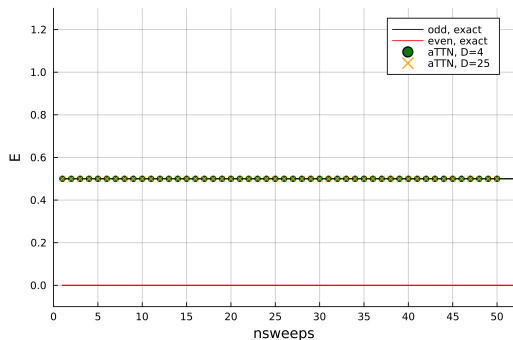
[Felser + *PRX* 10 (2020)]

[Felser *PhD thesis* (2021)]

Stability in the odd parity sector

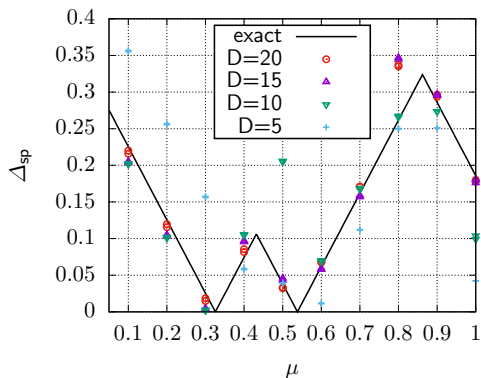


PEPS [Schneider, JO + *PRB* **104** (2021)]

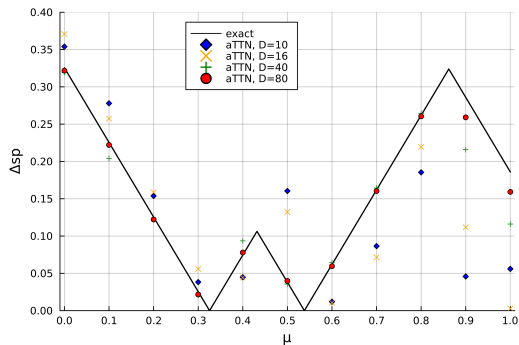


aTTN [Suladze, JO + (forthcoming)]

Simulations with chemical potential (3×4 , $U = 2$)

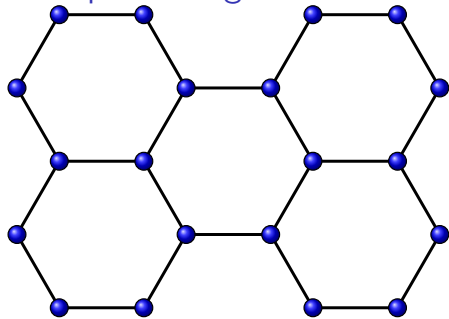


PEPS [Schneider, JO + *PRB* **104** (2021)]

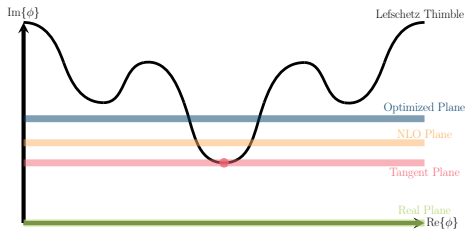
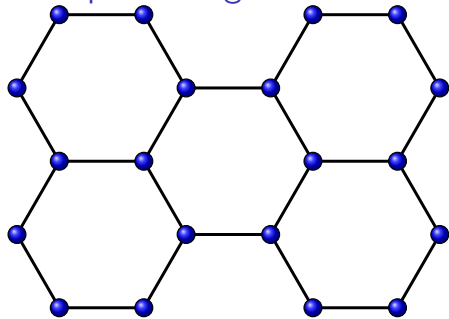


aTTN [Suladze, JO + (forthcoming)]

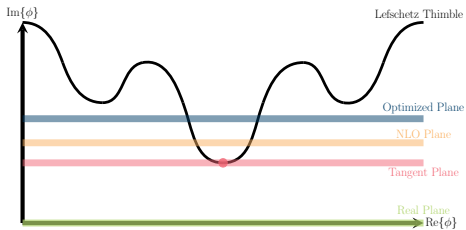
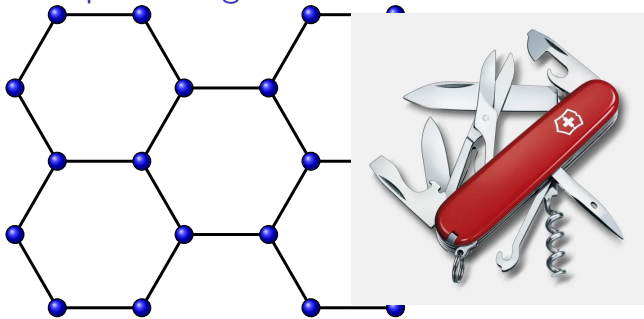
Constant path integral contour shifts



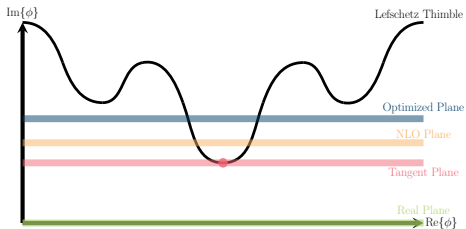
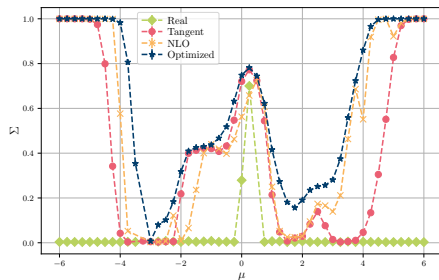
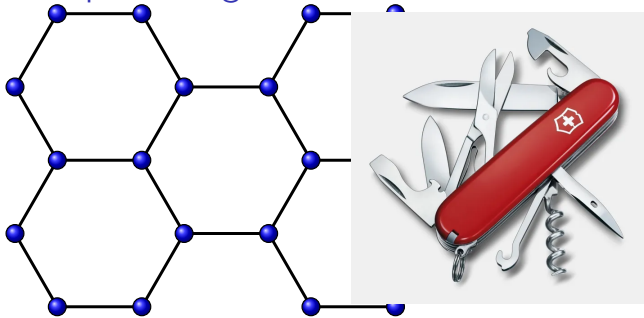
Constant path integral contour shifts



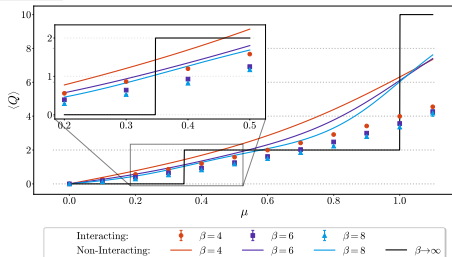
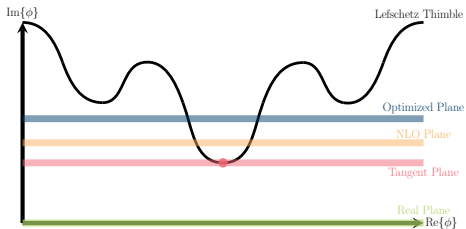
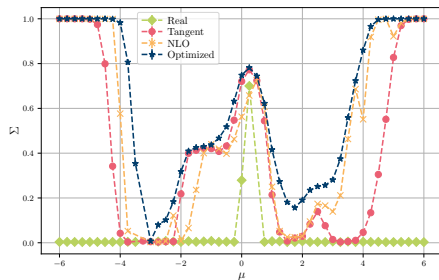
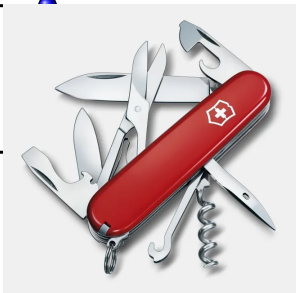
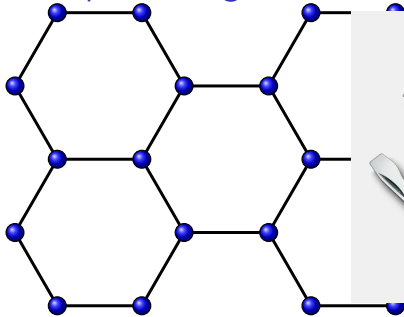
Constant path integral contour shifts



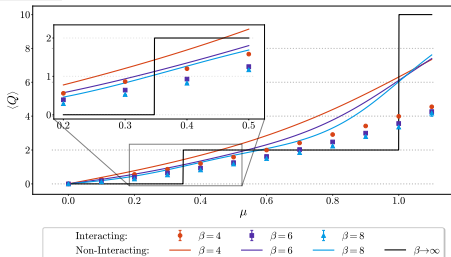
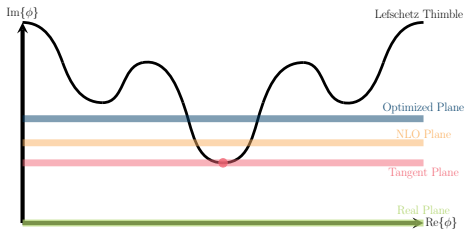
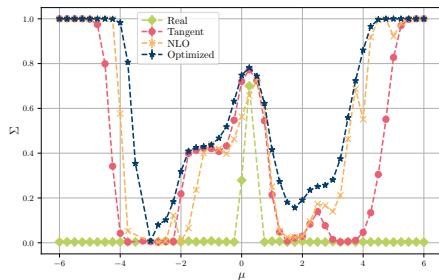
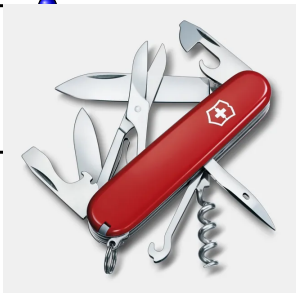
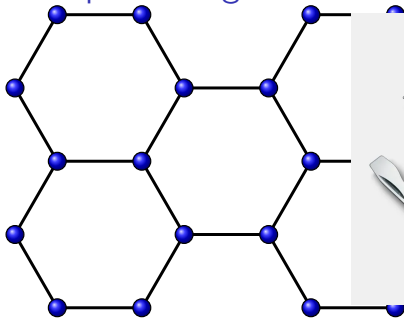
Constant path integral contour shifts



Constant path integral contour shifts



Constant path integral contour shifts



Stay tuned for updates on fermionic aTTNs!

parameters: dimension d , potential V with $V(x) \approx c|x|^a$ for large $|x|$

input : initial configuration $x^i \in X$, standard deviation σ

(default $\sigma = \sqrt{\frac{2}{ad}}$)

output : final configuration $x^f \in X$

sample $\gamma \sim \mathcal{N}(0, \sigma^2)$; // normal distribution

$x \leftarrow x^i \cdot e^\gamma$;

$\Delta V \leftarrow V(x) - V(x^i)$;

if $e^{-\Delta V + d\gamma} \geq \mathcal{U}_{[0,1]}$ **then** // uniform distribution

| $x^f \leftarrow x$;

else

| $x^f \leftarrow x^i$;

end

Path integral formalism

[Krieg, JO + CPC 236 (2019); Luu & Lähde PRB 93 (2016); Smith & Smekal PRB 89 (2014)]

- ▶ Discretise imaginary time into steps $\delta = \beta/N_t$, $\beta = 1/T$
- ▶ Hubbard-Stratonovich transformation

$$e^{-\frac{1}{2} \sum_{x,y} V_{x,y} q_x q_y} \propto \int \mathcal{D}\phi_t e^{-\frac{1}{2} \sum_{x,y} V_{x,y}^{-1} \phi_{x,t} \phi_{y,t} + i \sum_x \phi_{x,t} q_x}$$

- ▶ Fermion matrix

$$M_{(x,t)(y,t')} = \delta_{xy} \delta_{tt'} - e^{-i\delta \cdot \phi_{x,t}} \delta_{xy} \delta_{t-1,t'} - \delta \cdot \delta_{\langle x,y \rangle} \delta_{t-1,t'}$$

- ▶ Hybrid Monte Carlo simulation of probability

$$p[\phi] \propto \det(MM^\dagger) e^{-\frac{\delta}{2U} \phi^2}$$

Chemical potential

- ▶ $H_\mu = \mu \sum_x \left(c_{x,\uparrow}^\dagger c_{x,\uparrow} + c_{x,\downarrow}^\dagger c_{x,\downarrow} \right) = \mu \sum_x \left(p_x^\dagger p_x - h_x^\dagger h_x \right) + \text{const.}$
- ▶ μ introduces net charge in the ground state
- ▶ breaks particle-hole symmetry
- ▶ Probability weight

$$\det \left(M [\mu] M [-\mu]^\dagger \right) \not\equiv 0$$

→ Sign Problem

Statistical Power

$$\langle \mathcal{O} \rangle = \frac{\langle e^{-i\text{Im}S} \mathcal{O} \rangle_{\text{Re}S}}{\langle e^{-i\text{Im}S} \rangle_{\text{Re}S}} = \frac{1}{\Sigma} \langle e^{-i\text{Im}S} \mathcal{O} \rangle_{\text{Re}S}$$

$$\Sigma := \langle e^{-i\text{Im}S} \rangle_{\text{Re}S} \equiv \frac{\int \mathcal{D}\Phi e^{-\text{Re}S} e^{-i\text{Im}S}}{\int \mathcal{D}\Phi e^{-\text{Re}S}}$$

$$N^{\text{eff}} = |\Sigma|^2 \cdot N$$

$$\text{statistical error} \sim 1/\sqrt{N^{\text{eff}}} \propto 1/|\Sigma|$$

Transform path integral contour

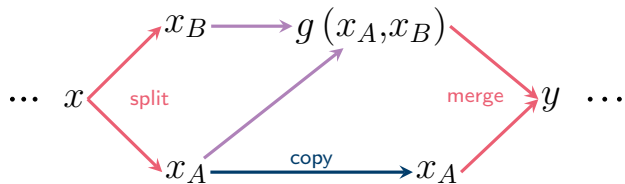
$$\begin{aligned} \mathcal{Z} &= \int_{\mathcal{M}} \mathcal{D}\Phi e^{-S[\Phi]} \\ &= \int_{\mathbb{R}^n} \mathcal{D}\phi \det J[\Phi(\phi)] e^{-S[\Phi(\phi)]}, \quad J_{ij} = \frac{\partial \Phi_i}{\partial \phi_j} \\ \Rightarrow S^{\text{eff}}[\phi] &= S[\Phi(\phi)] - \log \det J[\Phi(\phi)] \end{aligned}$$

Runtime of determinant as V^3 !

Affine coupling layers [Albergo + *hep-lat/2101.08176*; Dinh + *cs.LG/1410.8516*]

$$f(x) = \begin{cases} y_A = x_A \\ y_B = g(x_A, x_B) \end{cases}$$

$$g(x_A, x_B) = x_B \odot s(x_A) + t(x_A)$$



$$\Rightarrow \det \left(\frac{\partial f}{\partial x} \right) = \det \begin{pmatrix} \mathbb{1} & 0 \\ \frac{\partial y_B}{\partial x_A} & s(x_A) \end{pmatrix} = \prod_j s(x_A)_j$$

$$\det \frac{\partial \mathcal{N}\mathcal{N}}{\partial x} = \det \left(\frac{\partial f^n(x)}{\partial x} \right) \det \left(\frac{\partial f^{n-1}(x)}{\partial x} \right) \cdots \det \left(\frac{\partial f^1(x)}{\partial x} \right)$$

$$\det J = \det \left(\mathbb{1} + i \frac{\partial \mathcal{N}\mathcal{N}}{\partial x} \right) \neq \prod_k \det \left(\mathbb{1} + i \frac{\partial f^k(x)}{\partial x} \right)$$

Complex-valued Neural Networks

[Bassey + *stat.ML/2101.12249*; Bouboulis *cs.LG/1005.5170*; Brandwood *IET 130* (1983)]

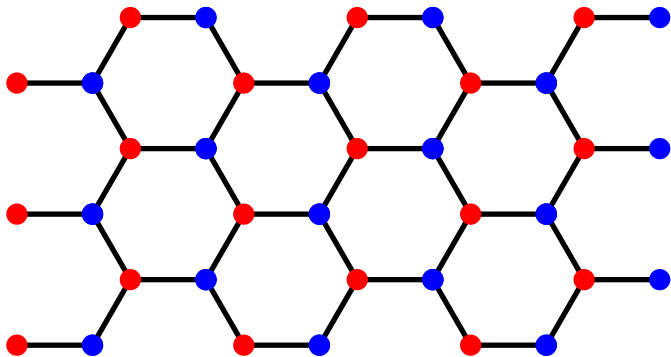
[Rodekamp, JO + *PRB 106* (2022)]

	real	complex
Trafo	$\phi \mapsto \phi + i\mathcal{NN}(\phi)$	$\phi \mapsto \mathcal{NN}(\phi)$
Derivative	$\frac{\partial f(x)}{\partial x}$	$\frac{\partial f(z)}{\partial z} = \frac{1}{2} \left(\frac{\partial f(z)}{\partial \text{Re}z} - i \frac{\partial f(z)}{\partial \text{Im}z} \right)$ $\frac{\partial f(z)}{\partial z^*} = \frac{1}{2} \left(\frac{\partial f(z)}{\partial \text{Re}z} + i \frac{\partial f(z)}{\partial \text{Im}z} \right)$
Layers	dense	affine
det J runtime	$\mathcal{O}(V^3)$	$\mathcal{O}(V)$

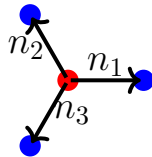
Tight binding model [Bloch 1929]

$$H^0 = - \sum_{\langle x,y \rangle} c_x^\dagger c_y$$

$\langle x,y \rangle$ denotes nearest neighbours

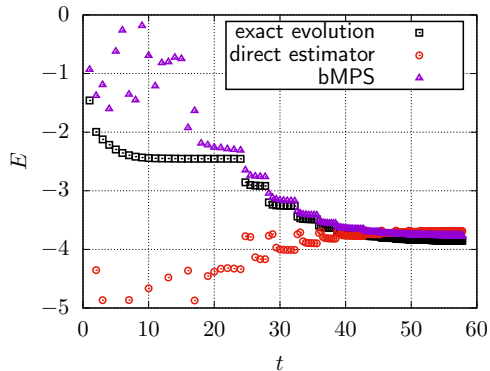


Sub-lattices **A** and **B**



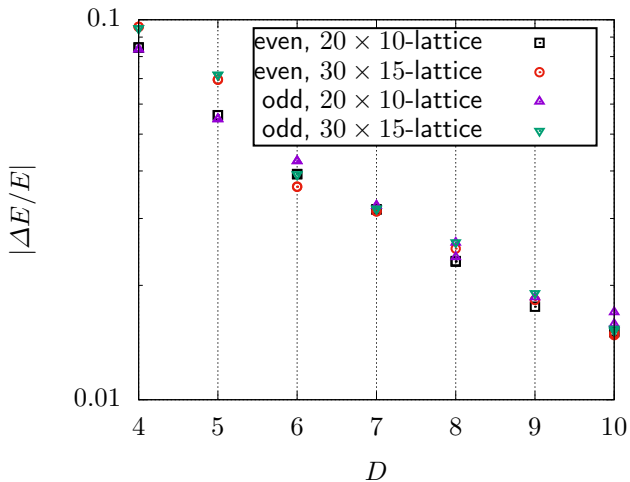
Ground state search [Schneider, JO + PRB 104 (2021)]

- ▶ Fix bond dimension D
- ▶ Initialise PEPS randomly
- ▶ Trotter-decomposed imaginary time evolution
- ▶ Local updates
- ▶ Contract network to calculate expectation values



Convergence (non-interacting $U = 0$, $\mu = 0.5$)

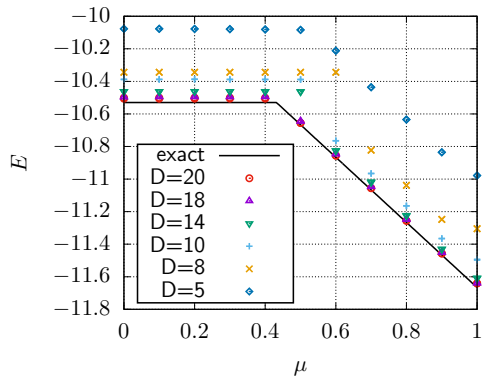
[Schneider, JO + *PRB* **104** (2021)]



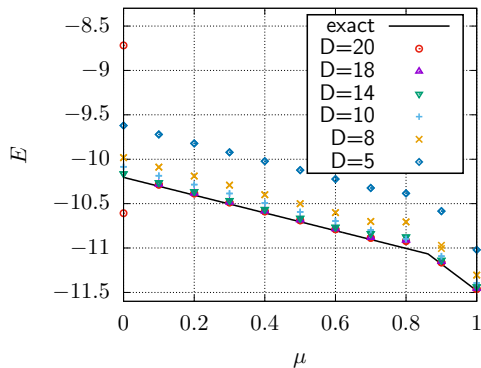
Simulations with chemical potential (3×4 , $U = 2$)

[Schneider, JO + PRB 104 (2021)]

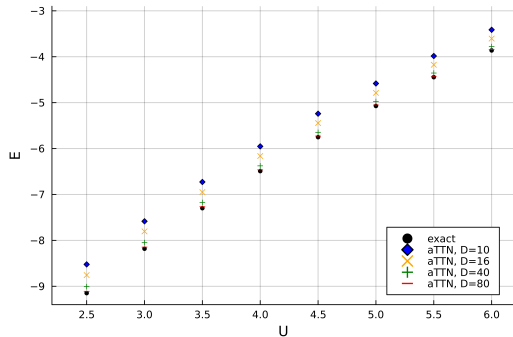
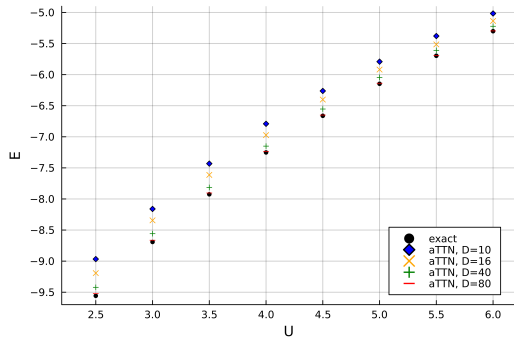
even parity



odd parity



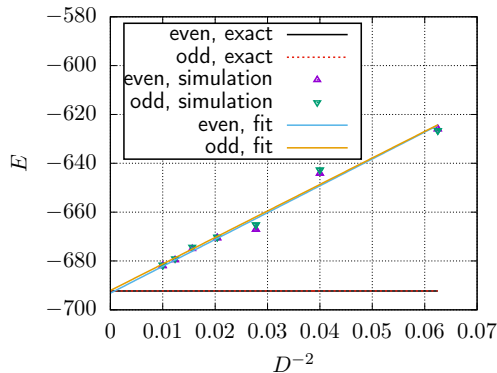
More preliminary fermionic aTTN results [Suladze, JO + (forthcoming)]



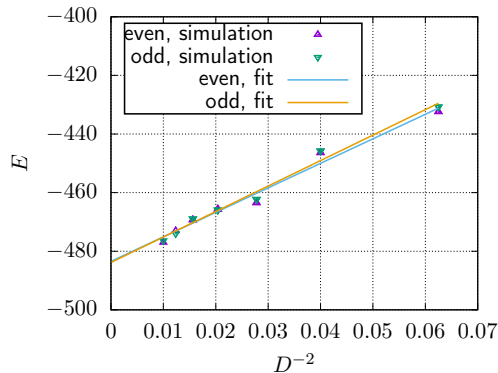
Simulations with chemical potential (30×15 , $\mu = 0.5$)

[Schneider, JO + PRB 104 (2021)]

$U = 0$



$U = 2$



Overview

	Hybrid Monte Carlo	Fermionic PEPS
lattice size	$L \lesssim 100$	$L \lesssim 30$
boundary conditions	periodic	open
thermodynamic limit	easy	hard
continuum limit	controlled, expensive	easy
temperature	only finite	only zero
other extrapolations	no	uncontrolled in D
excited states	few lowest, expensive	some specific, unstable
sign problem	yes	no
performance	CPU-intensive	RAM-intensive

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