

Machine Learning and Many-Body Physics



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Self-Learning quantum Monte Carlo method in interacting fermion systems

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IOP, CAS

7/7/2017

Serials works on SLMC

IOP: Zi Hong Liu, Zi Yang Meng

MIT: Junwei Liu, Yang Qi, Huitao Shen, Yuki Nagai, Liang Fu

UM: Kai Sun

arXiv:1610.03137

arXiv:1611.09364

arXiv:1612.03804

arXiv:1705.06724

arXiv:1706.10004

Similar work from Li Huang, Yi-feng Yang, Lei Wang

arXiv:1610.02746

arXiv:1612.01871

The power of SLMC

- **Reduce the time cost per sweep**
For $100 \times 100 \times 20$ lattice
DQMC ~ 150000 seconds/sweep
SLMC ~ 500 seconds/sweep
- **Reduce the auto-correlation time**
DQMC \sim may scale with system size at critical point
SLMC \sim constants ideally, model dependent

Determinantal QMC (DQMC)

DQMC(BSS algorithm)

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Monte Carlo calculations of coupled boson-fermion systems. I

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We present a formalism for carrying out Monte Carlo calculations of field theories with both boson and fermion degrees of freedom. The basic approach is to integrate out the fermion degrees of freedom and obtain an effective action for the boson fields to which standard Monte Carlo techniques can be applied. We study the structure of the effective action for a wide class of theories. We develop a procedure for making rapid calculations of the variation in the effective action due to local changes in the boson fields, which is essential for practical numerical calculations.

DQMC(BSS algorithm)

$$S = S_B + \int d\tau \int d^d x \psi^\dagger(x, \tau) \hat{O} \psi(x, \tau) \quad e^{-S_{\text{eff}}} = e^{-S_B} \det \hat{O}$$

$$D = \det_{x, \tau} \left(\frac{\partial}{\partial \tau} + H \right) \sim (\beta N)^3$$

$$= \det_x \left[I + T \exp \left(- \int_0^\beta d\tau H(\tau) \right) \right] \sim \beta N^3$$

DQMC basics

Slater determinant and its properties

Occupation number representation

N_e particle states $\hat{c}_1^\dagger \hat{c}_2^\dagger \cdots \hat{c}_{N_e}^\dagger |0\rangle$

operator with form $\hat{u} \equiv e^{-\hat{\mathbf{c}}^\dagger \mathbf{A} \hat{\mathbf{c}}}$ $\hat{u} \hat{c}_1^\dagger \hat{c}_2^\dagger \cdots \hat{c}_{N_e}^\dagger |0\rangle = \prod_{i=1}^{N_e} (\hat{\mathbf{c}}^\dagger e^{-\mathbf{A}})_i |0\rangle$

overlap of slater determinant

$$|\Psi\rangle = \prod_{i=1}^{N_e} (\hat{\mathbf{c}}^\dagger \mathbf{P})_i |0\rangle \quad |\tilde{\Psi}\rangle = \prod_{i=1}^{N_e} (\hat{\mathbf{c}}^\dagger \tilde{\mathbf{P}})_i |0\rangle$$

$$\langle \Psi | \tilde{\Psi} \rangle = \det [\mathbf{P}^\dagger \tilde{\mathbf{P}}]$$

DQMC basics

Hubbard-Stratonovich transformation

deal with interaction term

A continuous form

$$\exp\left(\frac{1}{2}\hat{A}^2\right) = \sqrt{2\pi} \int d\phi \exp\left(-\frac{1}{2}\phi^2 - \phi\hat{A}\right)$$

Other examples

$$\begin{aligned} e^{-\Delta\tau U(\hat{n}_\uparrow - 1/2)(\hat{n}_\downarrow - 1/2)} &= \frac{1}{2} e^{-\Delta\tau|U|/4} \sum_{s=\pm 1} e^{\alpha s(\hat{n}_\uparrow - \hat{n}_\downarrow)}, \quad U > 0 \\ &= \frac{1}{2} e^{-\Delta\tau|U|/4} \sum_{s=\pm 1} e^{\alpha s(\hat{n}_\uparrow + \hat{n}_\downarrow - 1)}, \quad U < 0 \end{aligned}$$

$$e^{\Delta\tau W \hat{A}^2} = \frac{1}{4} \sum_{l=\pm 2, \pm 1} \gamma(l) \exp\left(\sqrt{\Delta\tau W} \phi(l) \hat{A}\right) + o(\Delta\tau^4)$$

DQMC

Trotter decomposition

$$Z = \text{Tr} \left[e^{-\beta \hat{H}} \right] = \text{Tr} \left[\left(e^{-\Delta\tau \hat{H}_I} e^{-\Delta\tau \hat{H}_0} \right)^M \right] + \mathcal{O}(\Delta\tau^2)$$

HS transformation

$$Z = \sum_{\mathcal{C}} \mathcal{W}_{\mathcal{C}}^s \text{Tr} \left[\prod_{\tau=M}^1 e^{\hat{\mathbf{c}}^\dagger \mathbf{V}(\mathcal{C}) \hat{\mathbf{c}}} e^{-\Delta\tau \hat{\mathbf{c}}^\dagger \mathbf{T} \hat{\mathbf{c}}} \right] + \mathcal{O}(\Delta\tau^2)$$

Trace out fermions

(trace over all N_e particle basis, $N_e=1, \dots, N$)

$$Z = \sum_{\mathcal{C}} \mathcal{W}_{\mathcal{C}}^s \text{Tr} \left[\hat{U}(\beta, 0) \right] \quad \hat{U}(\tau_2, \tau_1) = \prod_{n=n_1+1}^{n_2} e^{\hat{\mathbf{c}}^\dagger \mathbf{V}(\mathcal{C}) \hat{\mathbf{c}}} e^{-\Delta\tau \hat{\mathbf{c}}^\dagger \mathbf{T} \hat{\mathbf{c}}}$$

$$Z = \sum_{\mathcal{C}} \mathcal{W}_{\mathcal{C}}^s \det [\mathbf{1} + \mathbf{B}(\beta, 0)] \quad \mathbf{B}(\tau_2, \tau_1) = \prod_{n=n_1+1}^{n_2} e^{\mathbf{V}(\mathcal{C})} e^{-\Delta\tau \mathbf{T}}$$

DQMC

partition function

$$Z = \sum_c \mathcal{W}_c^s \det [\mathbf{1} + \mathbf{B}(\beta, 0)]$$

Importance sampling of configurations

$$\begin{aligned} \langle \hat{O} \rangle &= \sum_c \mathcal{P}_c \langle \hat{O} \rangle_c \\ &\approx \frac{1}{N_{\text{sample}}} \sum_{i=1}^{N_{\text{sample}}} \langle \hat{O} \rangle_{c_i} \end{aligned}$$

$$\begin{aligned} \mathcal{P}_c &= \frac{\mathcal{W}_c^s \det [\mathbf{1} + \mathbf{B}(\beta, 0)]}{\sum_c \mathcal{W}_c^s \det [\mathbf{1} + \mathbf{B}(\beta, 0)]} \\ \langle \hat{O} \rangle_c &= \frac{\text{Tr} [\hat{U}(\beta, \tau) \hat{O} \hat{U}(\tau, 0)]}{\text{Tr} [\hat{U}(\beta, 0)]} \end{aligned}$$

Application of DQMC

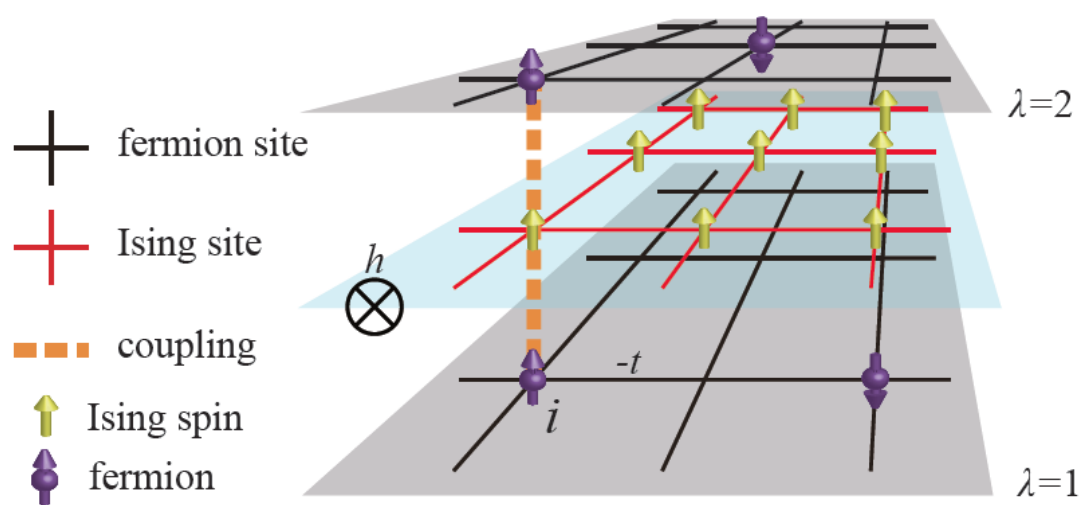
Coupled Fermion-boson lattice systems

- Interacting systems after HS transformation
Hubbard like models
Mott transition, chiral Ising and chiral Heisenberg transition, bosonic SPT etc.
- build-in coupled fermion-boson
AFM in metal, SDW, nematic QCP, FM QCP,
Z₂ deconfined phase transition

Issues of DQMC

- **Local update is level 1 BLAS algorithm**
Size limited, $L=20$ is the typical size
- **(critical) slowing down in some models**
build-in coupled fermion-boson problem

Slowing down for some models in DQMC

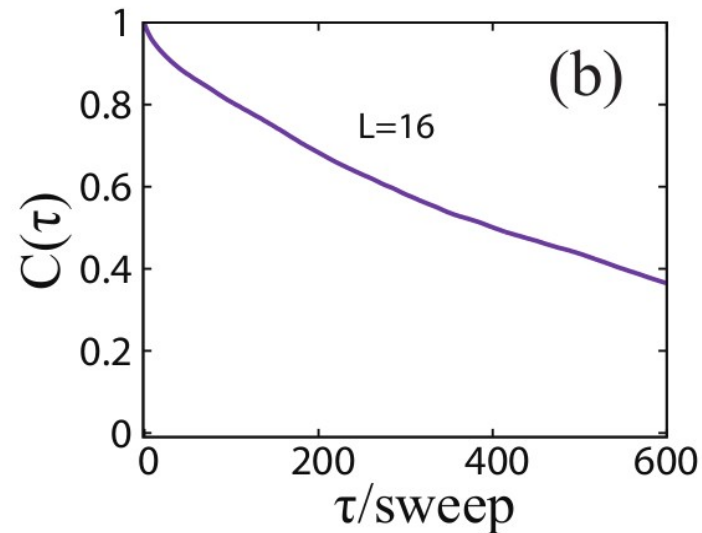
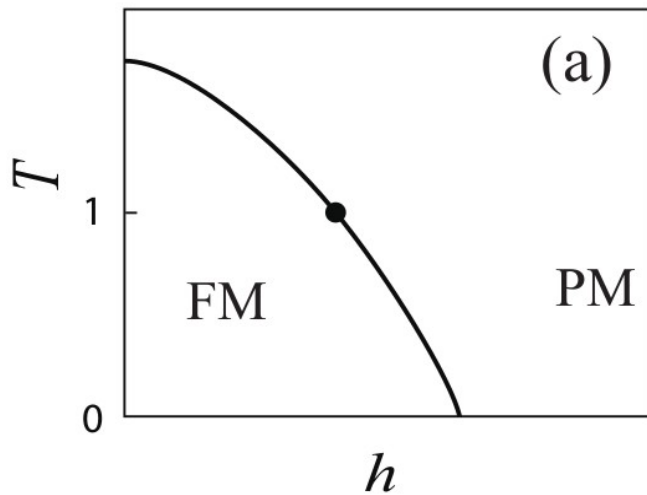


$$\hat{H} = \hat{H}_f + \hat{H}_s + \hat{H}_{sf}$$

$$\hat{H}_f = -t \sum_{\langle ij \rangle \lambda \sigma} \hat{c}_{i\lambda\sigma}^\dagger \hat{c}_{j\lambda\sigma} + h.c. - \mu \sum_{i\lambda\sigma} \hat{n}_{i\lambda\sigma}$$

$$\hat{H}_s = -J \sum_{\langle ij \rangle} \hat{s}_i^z \hat{s}_j^z - h \sum_i \hat{s}_i^x$$

$$\hat{H}_{sf} = -\xi \sum_i s_i^z (\hat{\sigma}_{i1}^z + \hat{\sigma}_{i2}^z)$$

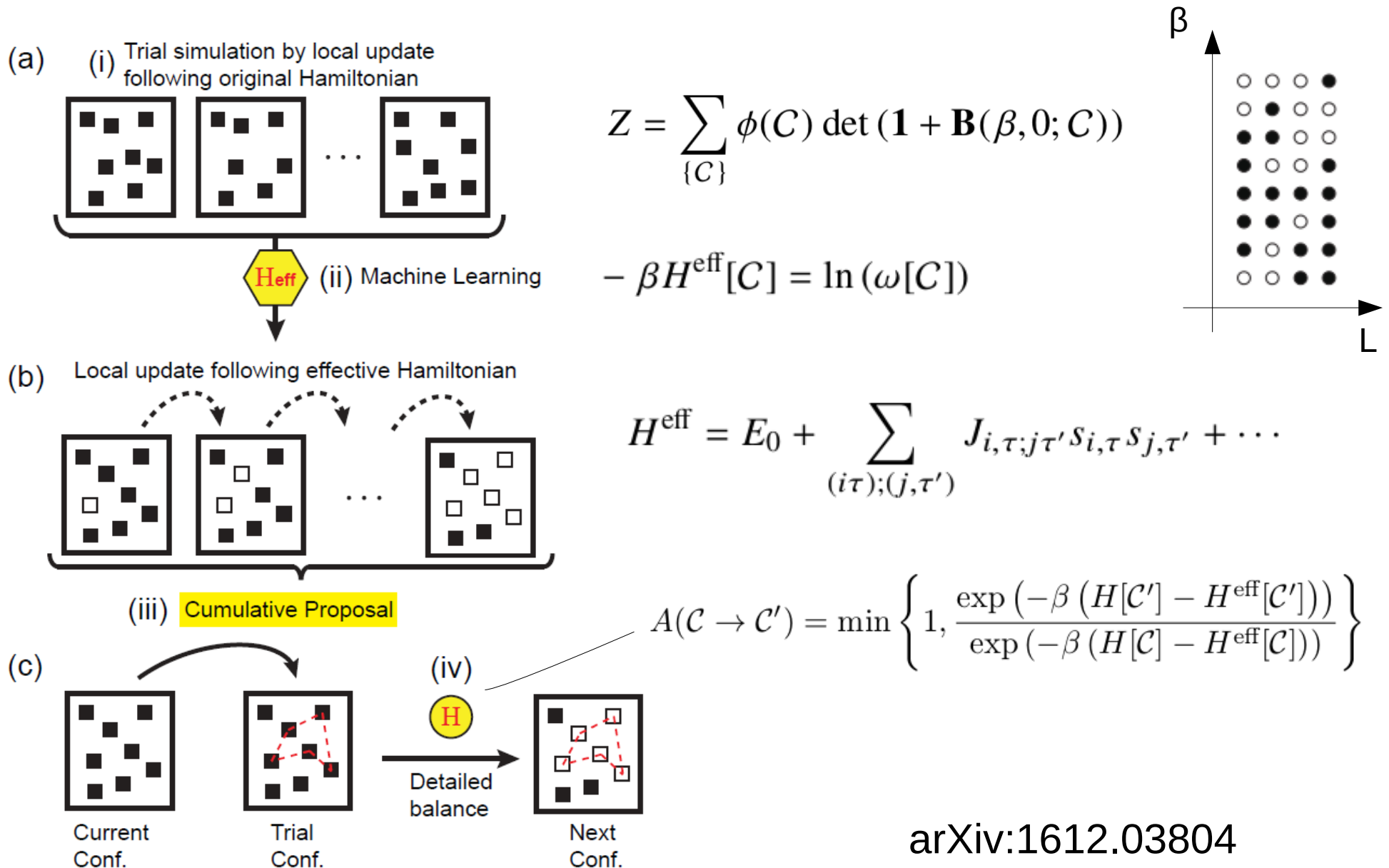


Complexity for getting an independent configuration: $\beta N^3 \tau_L$

Self-learning DQMC

arXiv:1612.03804
arXiv:1706.10004

Self-Learning Monte Carlo



Self-Learning Determinantal Quantum Monte Carlo

Complexity

- Cumulative update: $\gamma\beta N\tau_L$
- Detail balance: N^3

$$\begin{aligned}\omega_{\mathcal{C}} &= \phi(\mathcal{C}) \det(\mathbf{1} + \mathbf{B}(\beta, \tau)\mathbf{B}(\tau, 0)) \\ &= \phi(\mathcal{C}) \det(\mathbf{G}(0, 0))^{-1}\end{aligned}$$

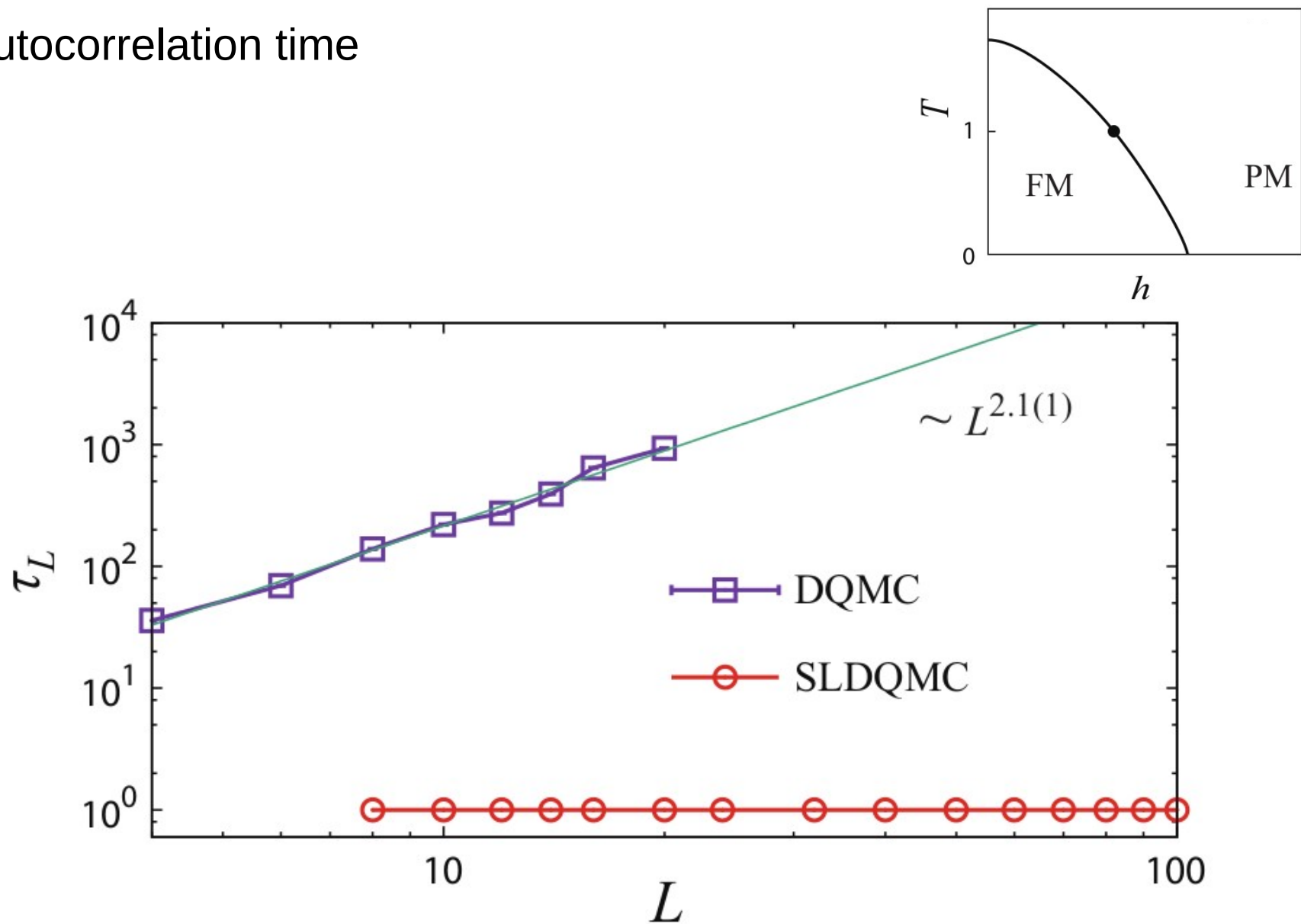
- Sweep Green's function: βN^2

$$\mathbf{G}(\tau + 1, \tau + 1) = \mathbf{B}(\tau + 1, \tau)\mathbf{G}(\tau, \tau)\mathbf{B}^{-1}(\tau + 1, \tau)$$

Complexity speed up $\mathcal{S} = \min\left(\frac{N^2}{\gamma}, N\tau_L, \beta\tau_L\right)$

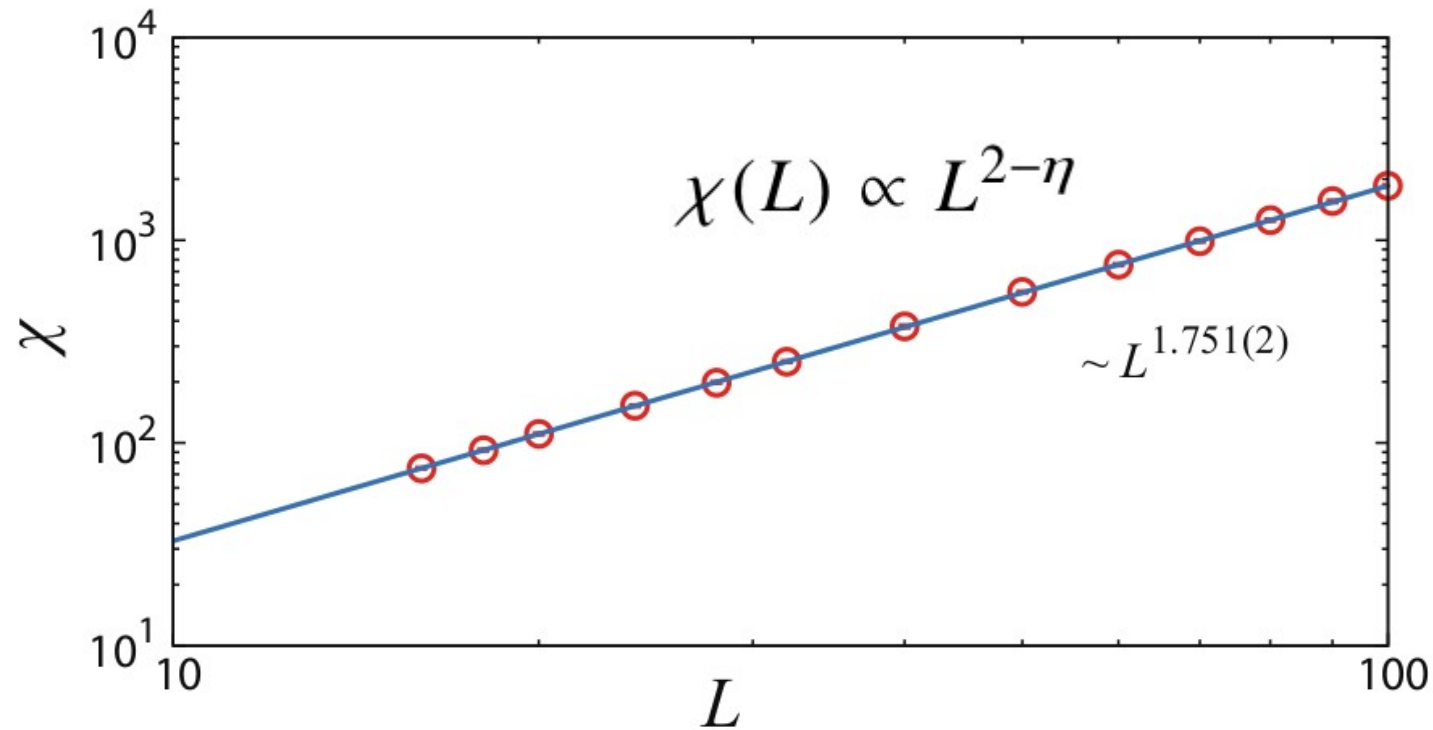
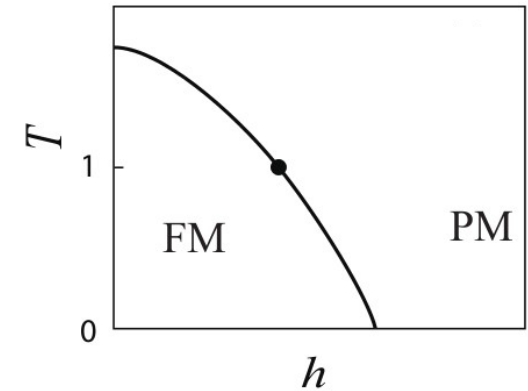
Self-Learning Determinantal Quantum Monte Carlo

Autocorrelation time



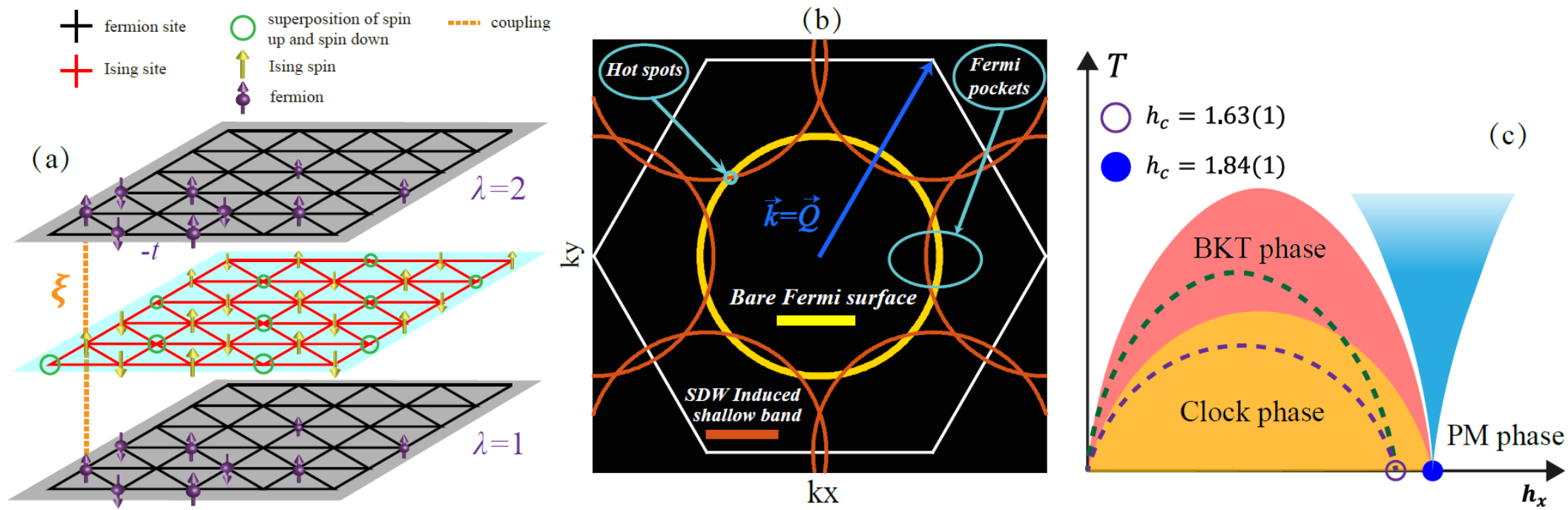
Self-Learning Determinantal Quantum Monte Carlo

$$\chi(L) = \frac{1}{L^2} \sum_{ij} \int_0^\beta d\tau \langle s_{i,\tau}^z s_{j,0}^z \rangle$$



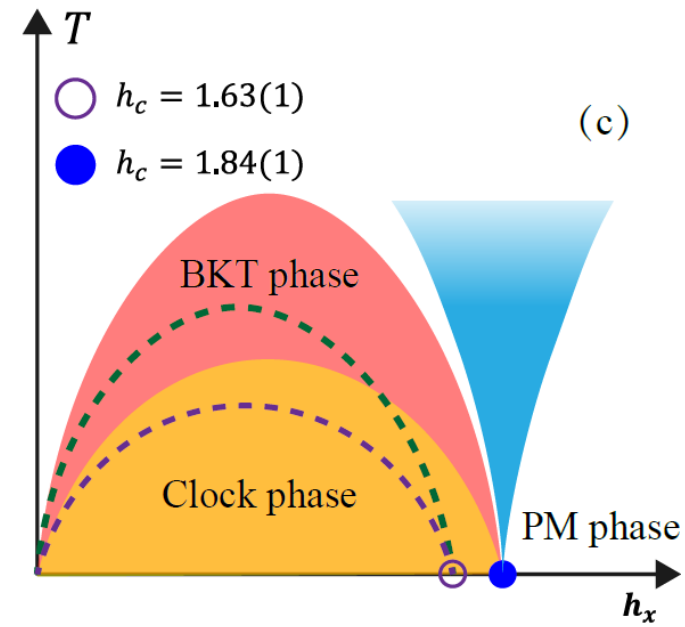
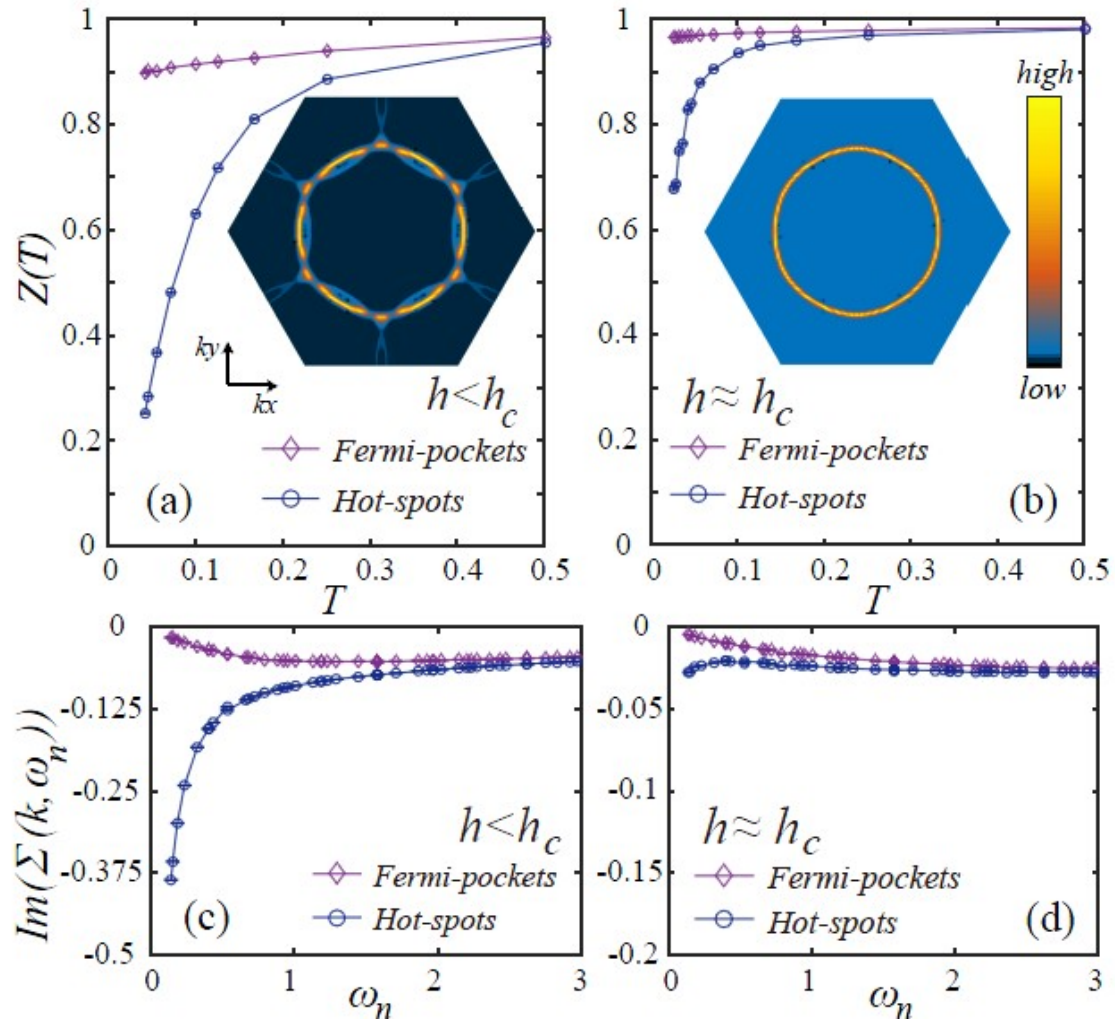
Using SLDQMC to attack hard problems

Itinerant quantum critical point with frustration



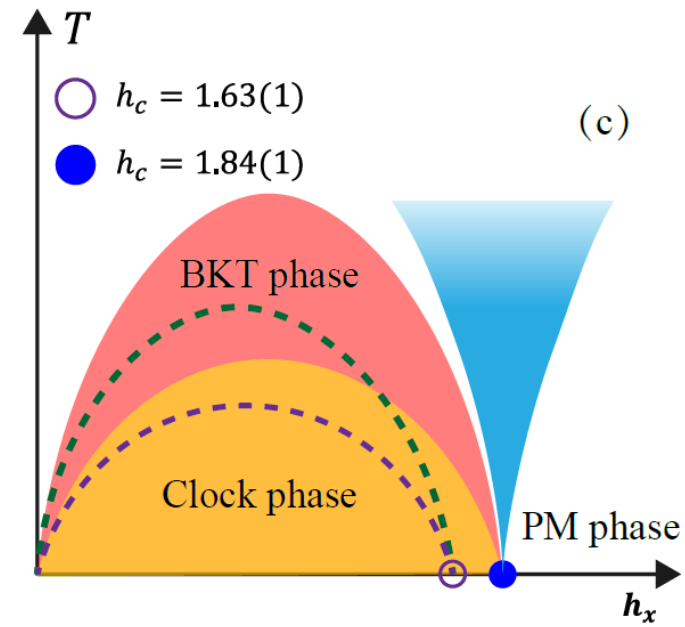
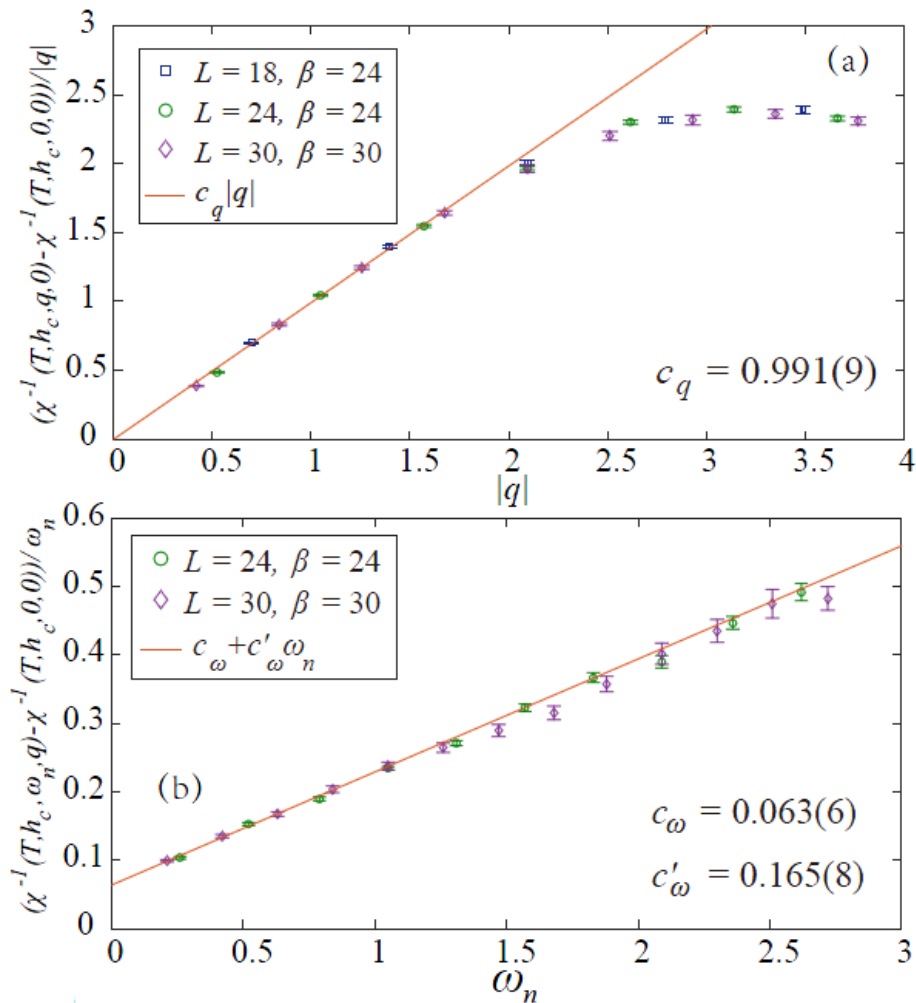
arXiv:1706.10004

Itinerant quantum critical point with frustration



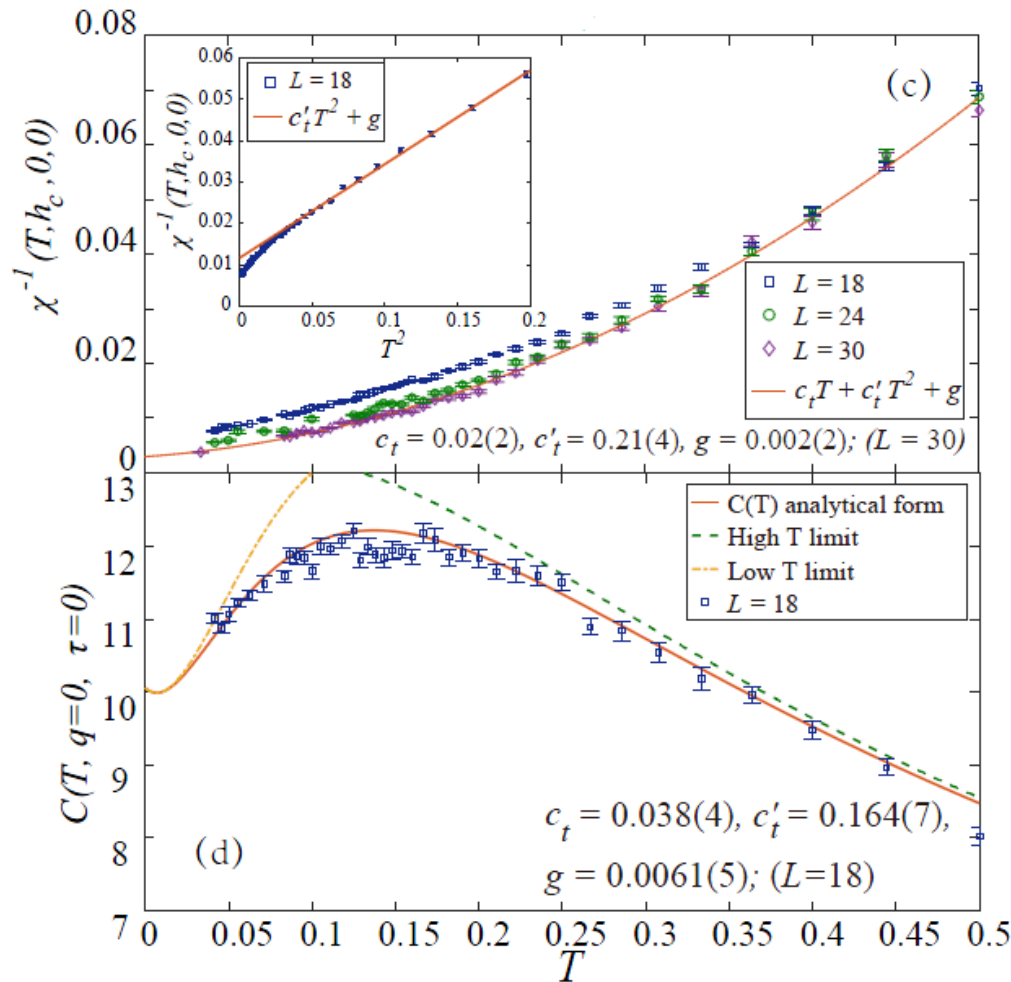
Non-fermi liquid behavior on hot spots

Itinerant quantum critical point with frustration

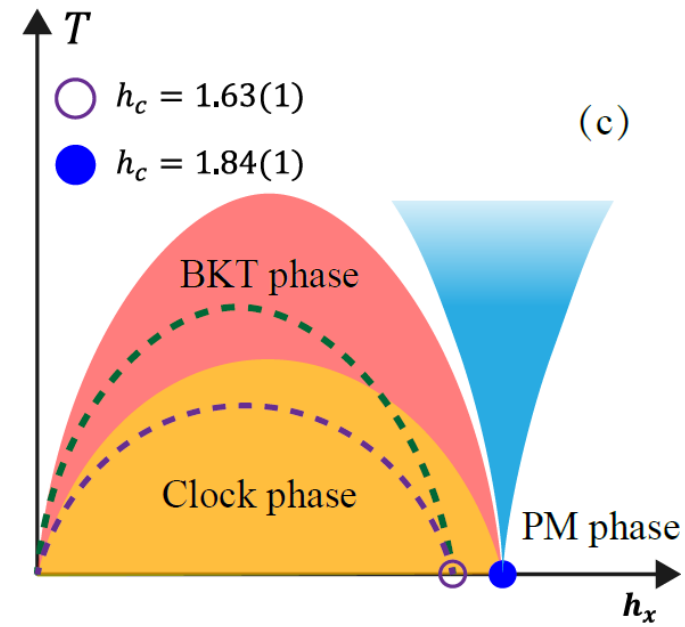


dynamical exponents $z=2$

Itinerant quantum critical point with frustration



$L=30$, $\beta=30$
($30 \times 30 \times 600$)



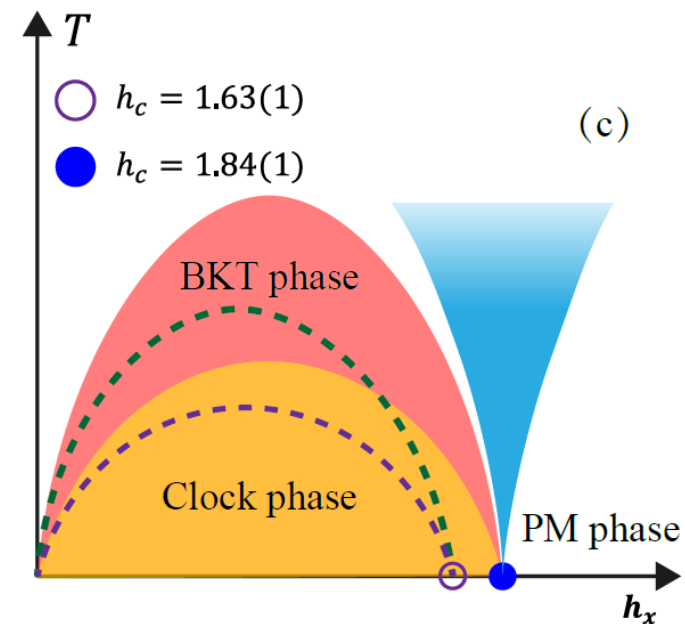
Linear T dependence in spin susceptibility

Itinerant quantum critical point with frustration

$$\chi(T, h, \mathbf{q}, \omega_n) =$$

$$\frac{1}{(c_t T + c'_t T^2) + c_h |h - h_c|^\gamma + c_q |\mathbf{q}|^2 + (c_\omega \omega + c'_\omega \omega^2)}$$

Hertz-Millis-Moriya theory on
finite momentum QCP



Summary and outlook

SLMC can be used to attack some hard problems

How general can it be is still a question

Self-Learning Monte Carlo

