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References

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Z. Y. Xie et al, PRL 103, 160601 (2009)
H.H. Zhao et al, PRB 81, 174411 (2010)
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H. H. Zhao et al, PRB 85, 134416 (2012)
Z. Y. Xie et al, arXiv:1201.1144

Why should we study the renormalization of tensors?

Strong correlated systems

Strong many-body effect:

- ✓ Single particle approximation invalid
- No good analytic tool: particles are highly entangled

Weak coupling systems

Week interaction

- Mean-field or single particle approximation works
- ✓ Particles are app. disentangled



Can we solve the problem numerically?



K. Wilson 1982



Walter Kohn 1997

Numerical Renormalization Group

Density Functional Theory



Three Stages of Numerical Renormalization Group Study

1. Wilson NRG 1975 -

0 Dimensional problems (single impurity Kondo model)

2. Density Matrix Renormalization Group (1D algorithm) 1992 -1 Dimensional quantum lattice models

3. Tensor Renormalization Group 2 or higher dimensional lattice models

Why is it more difficult to study higher-dimensional systems?



What is the wavefunction that satisfies this area law?

The Answer: Tensor Network State

$$|\Psi\rangle = Tr \prod T_{x_i x_i' y_i y_i'}[m_i]|m_i\rangle$$

Tensor network state (projected entangled pair state) is a faithful representation of the ground state of a quantum lattice model Verstraete, Cirac, arXiv:0407066

Classical Statistical Models = Tensor-network Model

$$Z = Tr \prod_{i} T_{x_i x_i' y_i y_i'}$$

The partitions for all statistical models with local interactions can be represented as tensor-network models



Tensor-Network Representation in the Original Lattice



Singular Value Decomposition



$$M_{S_iS_j} = \exp\left(-\beta H_{ij}\right) = \exp\left(\beta JS_iS_j\right)$$

$$M_{S_1S_2} = U_{S_1\sigma_1}A_{\sigma_1}U_{S_2\sigma_2}$$

Tensor-Network Representation in the Dual Lattice



$$H = -J \sum_{\langle ij \rangle} S_i S_j$$
$$Z = Tr \prod_{\Box} \exp(-\beta H_{\Box}) = Tr \prod_i T_{y_i x_i y'_i x'_i}$$
$$T_{\sigma_1 \sigma_2 \sigma_3 \sigma_4} = e^{-J\beta(\sigma_1 + \sigma_2 + \sigma_3 + \sigma_4)/2} \delta(\sigma_1 \sigma_2 \sigma_3 \sigma_4 - 1)$$



$$\sigma_{1} = S_{1}S_{2}$$

$$\sigma_{2} = S_{2}S_{3}$$

$$\sigma_{3} = S_{3}S_{4}$$

$$\sigma_{4} = S_{4}S_{1}$$

$$H_{\Box} = -J\left(\sigma_{1} + \sigma_{2} + \sigma_{3} + \sigma_{4}\right)/2$$

$$\sigma_{3}\sigma_{4} = S_{1}S_{2}S_{2}S_{3}S_{3}S_{4}S_{4}S_{1} = 1$$

Coarse Grain Tensor Renormalization Group (TRG)

Levin, Nave, PRL 99 (2007) 120601



Singular value decomposition: SVD best scheme for truncating a matrix

Accuracy of TRG

TRG is a good method, but it can be further improved



Ising model on a triangular lattice

Second renormalization of tensor-network state (SRG)

≻ TRG:

truncation error of *M* is minimized by the singular value decomposition

But, what really needs to be minimized is the error of Z!

➤ SRG:

The renormalization effect of M^{env} to M is considered

Xie et al, PRL **103**, 160601 (2009) Zhao, et al, PRB **81**, 174411 (2010)



I. Poor Man's SRG: entanglement mean-field approach

$$Z=Tr(MM^{env})$$

$$M_{kl,ij}^{env} pprox \Lambda_k^{1/2} \Lambda_l^{1/2} \Lambda_i^{1/2} \Lambda_j^{1/2}$$

Mean field (or cavity) approximation



Accuracy of Poor Man's SRG



Ising model on a triangular lattice

II. More accurate treatment of SRG

Evaluate the environment contribution *M*^{env} using TRG





1. Forward iteration

$$M^{(0)} \rightarrow M^{(1)}$$

 $\rightarrow \cdots \rightarrow M^{(N)}$

2. Backward iteration

$$M^{(N)} \to M^{(N-1)}$$

 $\to \dots \to M^{(0)} = M^{env}$

Accuracy of SRG



Ising model on a triangular lattice

Extension to 3D is difficult

Z. C. Gu, M. Levin, and X. G. Wen, unpublished



Efforts made by Nishino and collaborators

Corner Transfer Tensor Renormalization Group method Nishino & Okunishi, JPSJ 67, 3066 (1998)

Variational wavefunction (transfer matrix)

Okunishi, Nishino, Prog. Theor. Phys, 103, 541 (2000); Maeshima, Hieida, Akutsu, Nishino, Okunishi, PRE 64, 016705 (2001); Nishino, Hieida, Okunishi, Maeshima, Akutsu, Gendiar, Prog. Theo. Phys. 105, 409 (2001). Gendiar & Nishino, PRB 71, 024404 (2005)

> Main problem: tensor dimension D = 2 ~ 5 error > 0.6 %

A. Garcia-Saez, and J. I. Latorre, arXiv:1112.1412



TRG with Higher-Order Singular Value Decomposition of Tensors

Z. Y. Xie et al, arXiv:1201.1144



Higher-Order Singular Value Decomposition(HOSVD)

$$M_{xx'yy'}^{(n)} = \sum_{ijkl} S_{ijkl} U_{xi}^L U_{x'j}^R U_{yk}^U U_{y'l}^D$$



Nearly optimal low-rank approximation

L. de Latheauwer, B. de Moor, and J. Vandewalle, SIAM, J. Matrix Anal. Appl, 21, 1253 (2000).



Unitary Transformation Matrix

$$\varepsilon_1 = \sum_{i>D} |S(i,:,:,:)|^2$$

$$\varepsilon_2 = \sum_{j>D} |S(:,j,:,:)|^2$$

$$\bigcup^{\mathsf{U}}$$

$$\mathsf{U}^{\mathsf{U}}$$

 U^{D}

Only horizontal bonds need to be cut if $\varepsilon_1 < \varepsilon_2$, $U^{(n)} = U^L$ if $\varepsilon_1 > \varepsilon_2$, $U^{(n)} = U^R$

truncation error = min(ε_1 , ε_2)

$$T_{xx'yy'}^{(n+1)} = \sum_{ij} U_{ix}^{(n)} M_{ijyy'}^{(n)} U_{jx'}^{(n)}$$

How to do the HOSVD

HOSVD can be achieved by successive SVD for each index of the tensor

For example

$$M^{(n)}_{(x;x'yy')} = U^L \Lambda V^{\dagger}$$



Accuracy of HOTRG

Ising model on the square lattice



Hierarchical structure of HOTRG



Second Renormalization : HOSRG

Forward iterations: HOTRG to determine U⁽ⁿ⁾ and T⁽ⁿ⁾

backward iterations : evaluate the environment tensors



HOSRG: sweeping

Forward iterations:

Evaluate the bond density matrix

Find new $U^{(n)}$ and $T^{(n)}$



Accuracy of HOSRG

Ising model on the square lattice



HOSRG with forward-backward sweeping

Ising model on the square lattice





Three dimensions

Benchmark result for the 3D Ising model



 $T_c = 4.511546$



Relative Error

 $TRG < 10^{-6}$

other RG methods ~ 10⁻²

Thermodynamics of 2D QuantumTransverse Ising Model



Phase Transition with Partial Symmetry Breaking

QN Chen et al, PRL 107, 165701 (2011)



The Potts model is a basic model of statistical physics It has been intensively studied for more than 70 years

Full versus Partial Symmetry Breaking





full symmetry breaking

Entropy = 0

partial symmetry breaking

Entropy is finite



Ground States and Their Entropies

$$S = (N/2) \ln 2 + 2 * (3N/4) \ln \xi$$

- For a sublattice is ordered, the ground states are ξ^{3N/4}-fold degenerate $S = (3N/4) \ln \xi$
- both red and green sublattices are ordered, the ground states are 2^{N/2}-fold degenerate: S = (N/2) ln 2

Entropy and Partial Order



is ordered

 $\zeta = 1.7525 > 2^{2/3} \approx 1.5874$

Conjecture: There is a Finite T Phase Transition



There is a finite T phase transition with two singularities:

- 1. ordered and disordered states
- 2. Z_2 between green and red

Phase Transition: Specific Heat Jump



Partial Order Phase Transition in Other Irregular Lattices



Summary

- 1. HOSRG provides an accurate and efficient numerical method for studying 2D or 3D classical/quanutum lattice models
- 2. AF Potts Model has an entropy driven partial ordering and finite T phase transition on irregular lattices