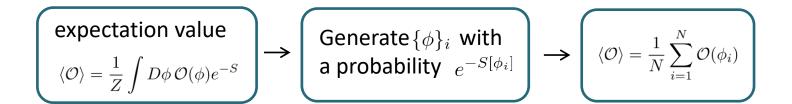
Tensor network methods in four dimensional field theory

Daisuke Kadoh NCTS, National Tsing-Hua Univ.

Numstrings 2021 (online) Jan 20, 2021

MC method and the sign problem

Monte Carlo method

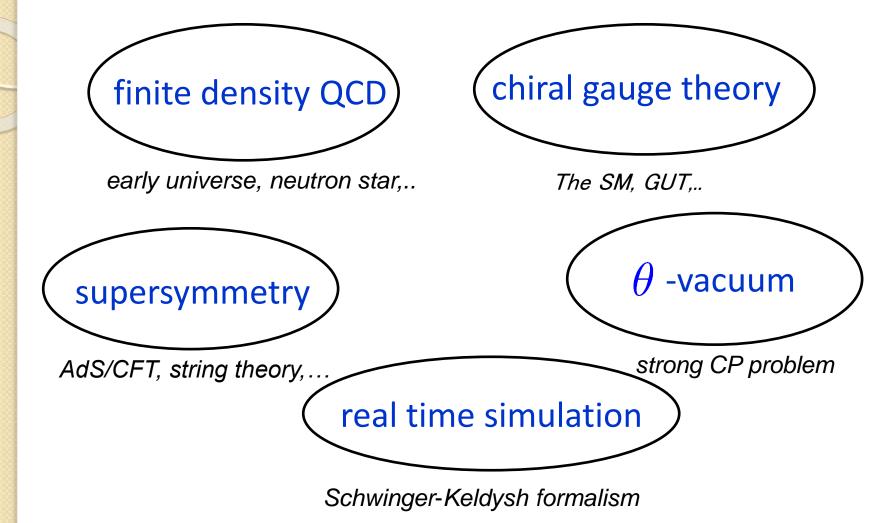


 $e^{-S[\phi_i]}$: positive real number

(example) SYM QM

$$S = \frac{N}{2\lambda} \int dt \, \mathrm{tr} \left\{ (D_0 X_i)^2 - \frac{1}{2} [X_i, X_j]^2 + \psi D_0 \psi + \psi \gamma_i [X_i, \psi] \right\}$$
 $Z = \int DADX e^{-S_B} \mathrm{pf}(D)$ negative can be negative sign problem

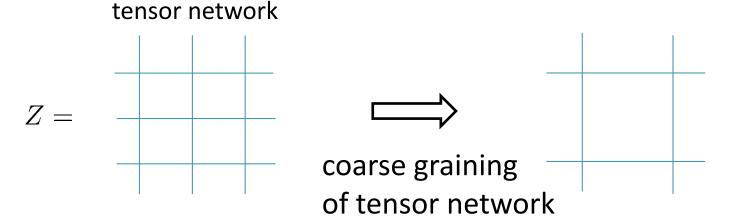
Cases with the sign problem



It is important to solve the sign problem.

Tensor renormalization group (TRG)

Levin-Nave, 2007



no stochastic process & no sign problem

Previous studies in field theory:

- Y. Kuramashi, S.Takeda, Y. Meurice, K.Jansen, M.C.Banuls, D.C-J.Lin,
- S. Catterall, J. Unmuth-Yockey, A.Bazavov, R.Sakai, Y. Yoshimura,
- R.G.Jha, H. Oba, S. Akiyama, ...

c.f. improved MC method: reweighting method, complex Langevin, Lefschetz thimble, ...

Talk plan

- 1. Motivation
- 2. TRG in 2d Ising model and scalar field theory
- 3. The triad TRG method
- 4. Summary and future outlook

2. TRG in 2d Ising model and scalar field theory

2d Ising model

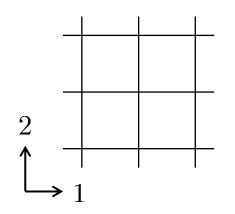
Hamiltonian

$$H = -J\sum_{n} \{\sigma_{n}\sigma_{n-\hat{1}} + \sigma_{n}\sigma_{n-\hat{2}}\} + h\sum_{n} \sigma_{n}$$

$$\sigma_n = \pm 1$$

$$n = (n_1, n_2), \quad n_i \in \mathbf{Z}$$

$$\hat{\mu} \quad \text{: unit vector of } \mu\text{-direction}$$

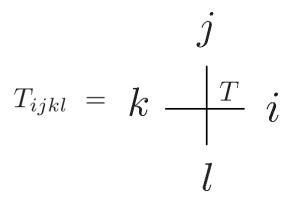


Partition function

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

$$\equiv \prod_{n} \sum_{\sigma_n = \pm 1} e^{\beta J(\sigma_n \sigma_{n-\hat{1}} + \sigma_n \sigma_{n-\hat{2}}) + \beta h \sigma_n}$$

Graphical representation of tensors

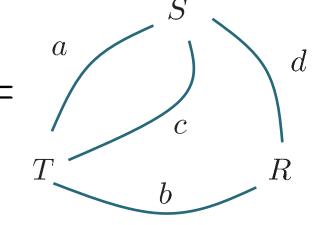


"contraction of tensors"

$$\sum_{k=1}^{N} T_{kabc} S_{ijk} = a \frac{T}{c} \frac{1}{k}$$

"tensor network"

$$Z = \sum_{a,b,c,d} T_{bca} S_{acd} R_{bd}$$



The TN representation of Z

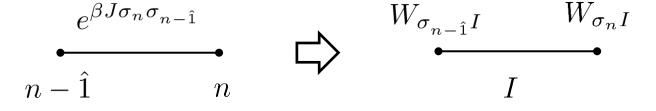
(1) decompose the hopping term

$$e^{\beta J \sigma_n \sigma_{n-\hat{1}}} = \cosh(\beta J) + \sigma_n \sigma_{n-\hat{1}} \sinh(\beta J)$$

$$= \sum_{I=0,1} \cosh(\beta J) (\sigma_n \sigma_{n-\hat{1}} \tanh(\beta J))^I$$

$$= \sum_{I=0,1} W_{\sigma_n I} W_{\sigma_{n-\hat{1}} I}$$

$$W_{\sigma I} = \sqrt{\cosh(\beta J)} \sqrt{\tanh(\beta J)}^{I} \sigma^{I}$$



The TN representation of Z (cont'd)

(2) make tensor from $W_{\sigma_n I}$

$$Z = \sum_{I=0,1} \sum_{J=0,1} \sum_{K=0,1} \sum_{L=0,1} \cdots W_{\sigma_n J}$$

$$\times \left\{ \sum_{\sigma_n = \pm 1} W_{\sigma_n I} W_{\sigma_n J} W_{\sigma_n K} W_{\sigma_n L} \right\} \cdots$$

$$= \sum_{I=0,1} \sum_{J=0,1} \sum_{K=0,1} \sum_{L=0,1} \cdots T_{IJKL} \cdots$$

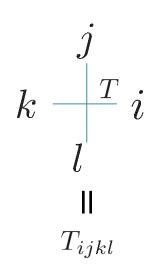
$$T_{IJKL} = \sum_{\sigma=-1,+1} W_{\sigma I} W_{\sigma J} W_{\sigma K} W_{\sigma L}$$

$$I \longrightarrow K$$

The TN representation of Z (cont'd)

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

$$= \sum_{l,o,\dots} T_{ijkl} T_{lmno} T_{opqr} \cdots$$



Lattice field theory (translational invariance)

→ local and homogeneous tensor networks

Singular value decomposition (SVD)

SVD of N x N matrix T_{IJ}

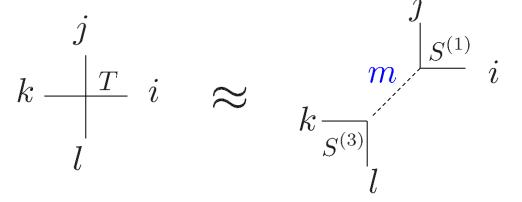
$$T_{IJ} = \sum_{m=1}^{N} U_{Im} \sigma_m V_{mJ}$$
 singluar values $\sigma_1 \geq \sigma_2 \geq \cdots \geq \sigma_N \geq 0$ $= \sum_{m=1}^{N} S_{Im} S'_{mJ}$ $S_{Im} = \sqrt{\sigma_m} U_{Im}$ $S'_{mJ} = \sqrt{\sigma_m} V_{mJ}$ $I \longrightarrow J = I \longrightarrow J$

low rank approximation

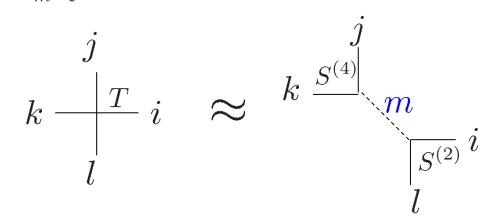
$$T_{IJ} pprox \sum_{m=1}^{D_{\mathrm{cut}}} S_{Im} S'_{mJ}$$
 $D_{cut} < N$

SVD for tensors

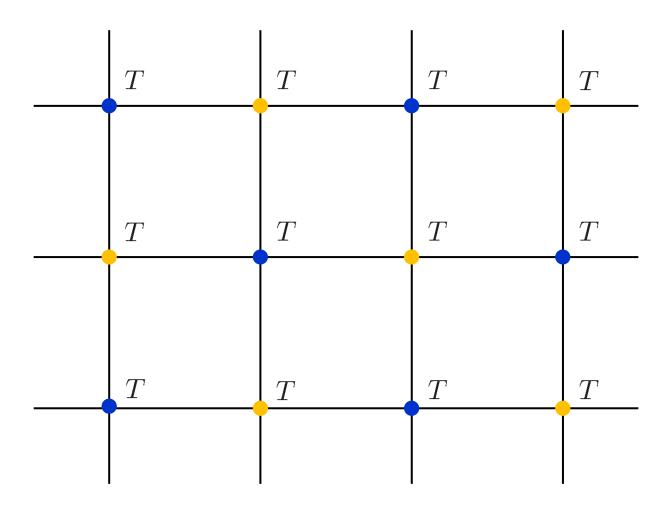
• $T_{ijkl} pprox \sum_{m=1}^{D_{cut}} S_{ijm}^{(1)} S_{klm}^{(3)}$ on even sites ($n_1+n_2= ext{even}$)



• $T_{ijkl} \approx \sum_{m=1}^{D_{cut}} S_{lim}^{(2)} S_{jkm}^{(4)}$ on odd sites ($n_1 + n_2 = \text{odd}$)



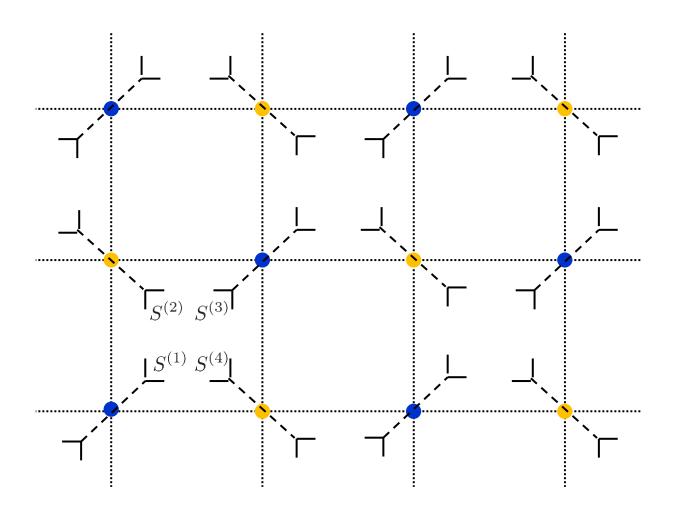
Renormalization step



even

odd

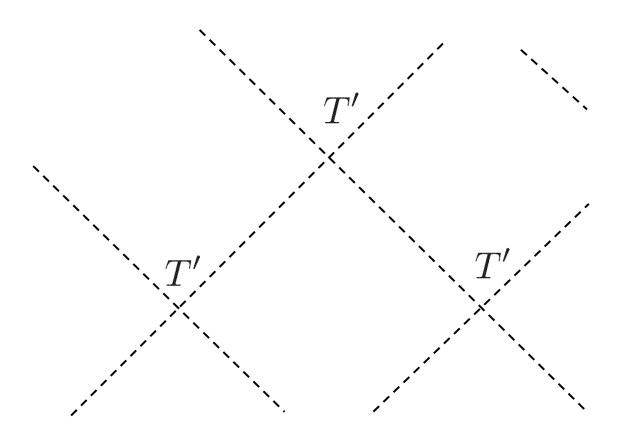
Renormalization step



even

odd

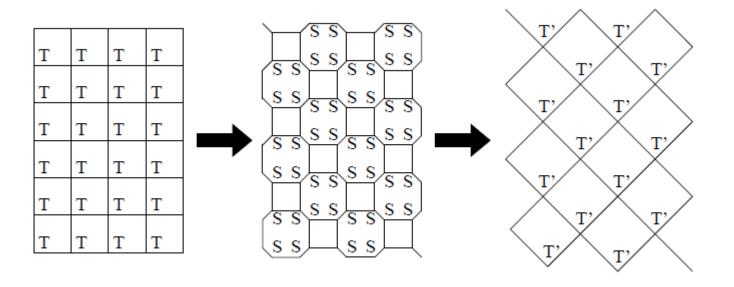
Renormalization step



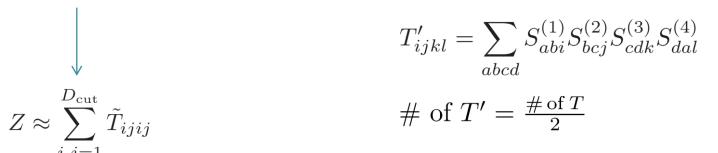
$$T'_{ijkl} = \sum_{abcd} S^{(1)}_{abi} S^{(2)}_{bcj} S^{(3)}_{cdk} S^{(4)}_{dal}$$

Tensor renormalization group

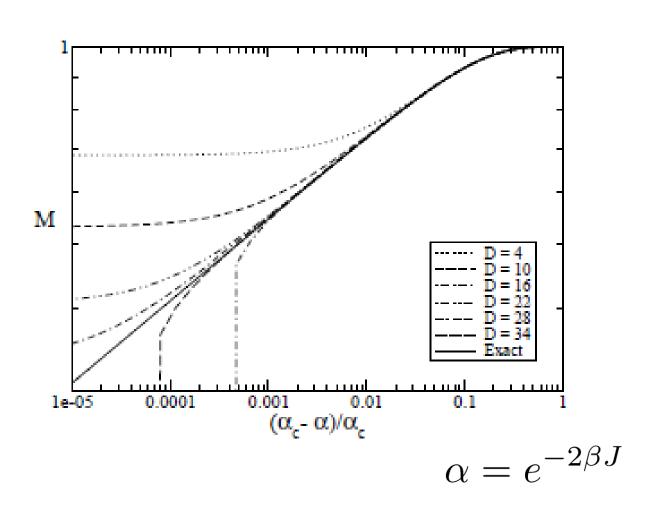
Levin-Nave, 2007



$$Z = \sum_{i,j,...} T_{ijkl} T_{lmno} \cdots \approx \sum_{i,j,...} S^{(1)} S^{(2)} S^{(3)} S^{(4)} \cdots = \sum_{m,n,...} T'_{mnpq} T'_{mkrs} \cdots$$



Levin-Nave, 2007



Properties of TRG

- (1) no sign problem, no statistical errors
- (2) systematic error from finite Dcut
- (3) partition function is directly calculable
- (4) computational cost is $\log(V) \times D_{cut}^p$

$$p = 5 \text{ for } d = 2 \text{ (TRG,2007)}$$

$$p = 4d - 1$$
 (HOTRG,2012)

$$p = 2d + 1$$
 (ATRG,2019)

2d complex ϕ^4 -theory at finite density

continuum action

$$S = \int d^2x \, \left\{ |\nabla_\rho \phi|^2 + (m^2 - \mu^2) |\phi|^2 + \mu (\partial_0 \phi^* \phi - \phi^* \partial_0 \phi) + \lambda \phi^4 \right\}$$

$$m^2, \lambda > 0 \qquad \text{mass and coupling constant}$$

$$\mu \geq 0 \qquad \text{chemical potential}$$

sign problem at finite chemical potential

$$S[\mu]^* = S[-\mu]$$

• bulk observables are independent of the chemical potential for $\mu < \mu_c$ (Silver Blaze phenomenon)

Discretization of scalar fields

Gaussian quadrature

$$\int_{-\infty}^{\infty} d\phi F(\phi) \approx \sum_{\phi \in S_K} g_K(\phi) F(\phi)$$

e.g. Gauss-Hermite quadrature

$$S_K$$
 = a set of roots of $H_K(\phi)$
$$g_K(\phi) = \frac{2^{K-1}K!\sqrt{\pi}}{K^2H_{K-1}^2(\phi)}e^{\phi^2}$$

discretization of the path integral measure

$$\int d\phi_1 d\phi_2 \cdots \phi_N \approx \sum_{\phi_1 \in S_K} \sum_{\phi_2 \in S_K} \cdots \sum_{\phi_N \in S_K} g_K(\phi_1) g_K(\phi_2) \cdots g_K(\phi_N)$$

TN representation of scalar theory

decompose the hopping term

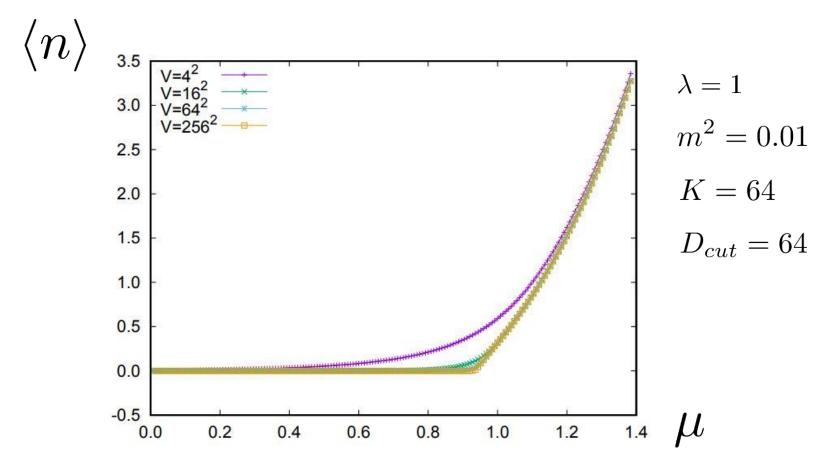
$$\begin{split} h(\phi_{n},\phi_{n-\hat{1}}) &= \exp\left\{-\frac{1}{2}(\phi_{n}-\phi_{n-\hat{1}})^{2} - \frac{1}{8}\mu_{0}^{2}(\phi_{n}^{2}+\phi_{n-\hat{1}}^{2}) - \frac{\lambda_{0}}{16}(\phi_{n}^{4}+\phi_{n-\hat{1}}^{4})\right\} \\ &= \sum_{I=1}^{K^{2}} U_{\phi_{n}I}\sigma_{I}V_{I\phi_{n-\hat{1}}}^{\dagger} \qquad \text{(SVD of KxK matrix)} \\ &= \sum_{I=1}^{K^{2}} W_{\phi_{n}I}^{(1)}W_{\phi_{n-\hat{1}}I}^{(2)} \\ &W_{\phi I}^{(1)} = U_{\phi I}\sqrt{\sigma_{I}} \qquad W_{\phi I}^{(2)} = \sqrt{\sigma_{I}}V_{I\phi}^{\dagger} \end{split}$$

make tensor from W:

$$T_{IJKL} = \sum_{\phi \in S_K} g_K(\phi) W_{\phi I}^{(1)} W_{\phi J}^{(1)} W_{\phi K}^{(2)} W_{\phi L}^{(2)}$$
$$I, J, K, L = 1, 2, \dots, D$$

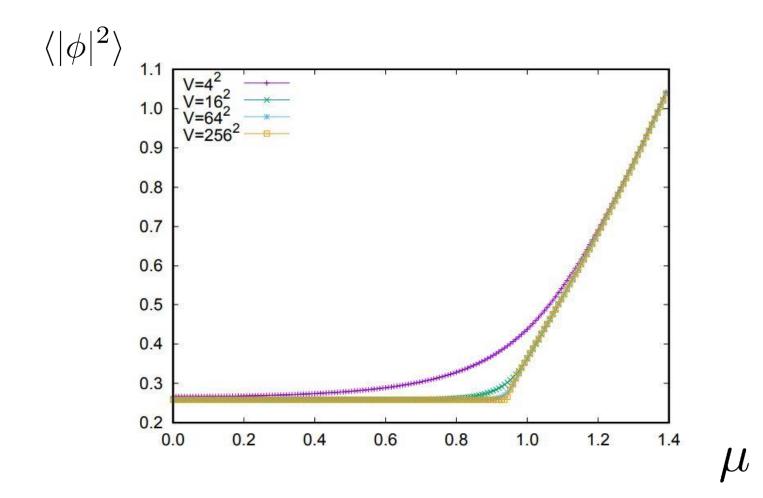
The particle number density

[D.K., Kuramashi, Nakamura, Sakai, Takeda, Yoshimura (2019)]



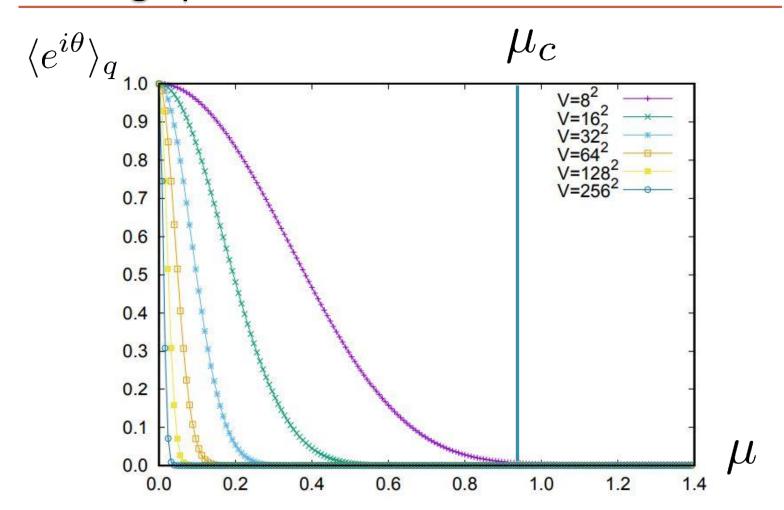
The Silver Blaze is clearly observed for large volume lattices.

The expectation value of $|\phi|^2$



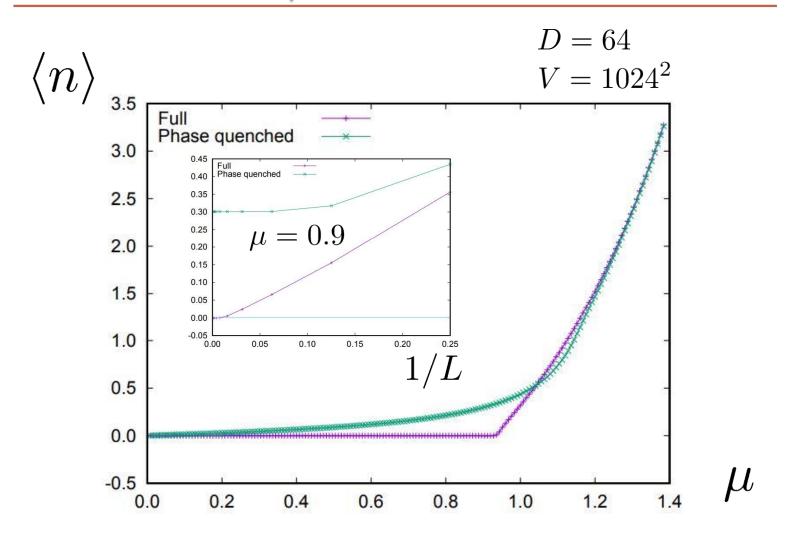
Results similar to number density can be seen.

Average phase factor



The sign problem becomes severe for large volume lattices.

Silver Blaze vs. phase



The Silver braze is not obtained without the phase, and TRG properly works for LFT with severe sign problem.

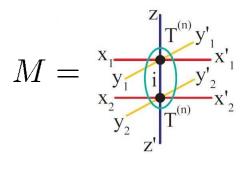
3. The triad TRG method

Kadoh and Nakayama, arXiv:1912.02414

Higher-order TRG (HOTRG)

3d case

Xie et al., 2012



(1) Make projectors from two Ts

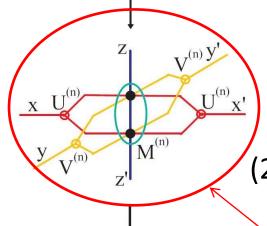
$$M = T \cdot T$$

$$\mathcal{O}(D^{2d+2})$$

diagonalization:

$$(UMM)U^{\dagger})_{XX'} = \sigma_X \delta_{XX'}$$

 U_{X,x_1x_2} : projector for x-direction



(2) Take contractions with projectors
 → a renormalized tensor T'

$$T' = \sum_{\mathbf{y} \in \mathbf{T}^{(n+1)}}^{\mathbf{z}} \mathbf{y}'$$

$$\mathcal{O}(D^{4d-1})$$

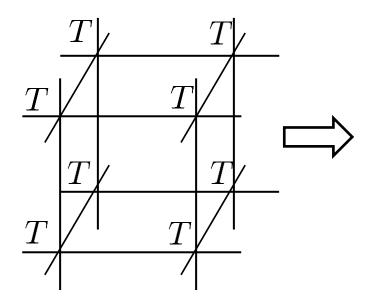
Can we wait?

computational time is a few hours for D=32 in 2d. however, in 4d, it becomes...

$$1.5[h] \times 32^8 \; \rightleftharpoons \; 200 \; \text{million years!!!}$$
 $D^7 \longleftrightarrow \; D^{15}$

→ need to create a low-cost scheme applicable to higher dimensions

Why HOTRG's cost is high?



T =

2d-rank tensor

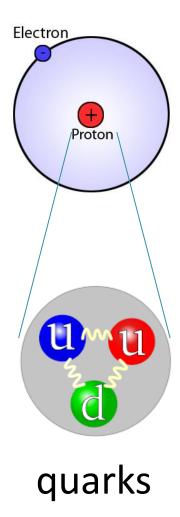


tensor networks on hyper cubic lattice

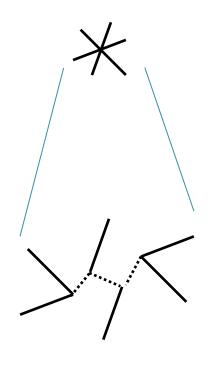
The cost of contracting two 2d-rank tensors is high for large d.

We should reconsider a theory of tensor networks at a fundamental level.

Fundamental building blocks



rank-6 tensor



rank-3 tensors

"triads"

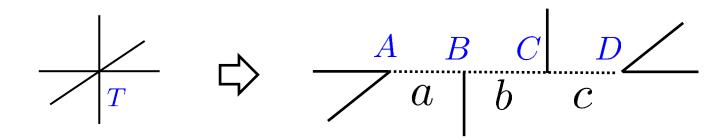
Hidden structure

tensor in the polyadic decomposition

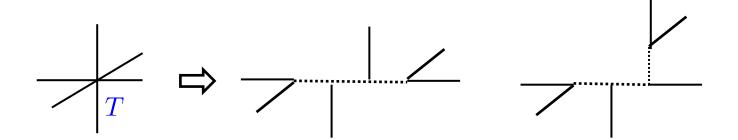
$$T_{ijklmn} = \sum_{a=1}^{r} W_{ia}^{(1)} W_{ja}^{(2)} W_{ka}^{(3)} W_{la}^{(4)} W_{ma}^{(5)} W_{na}^{(6)}$$
$$= \sum_{a,b,c=1}^{r} A_{ija} B_{akb} C_{blc} D_{cmn}$$

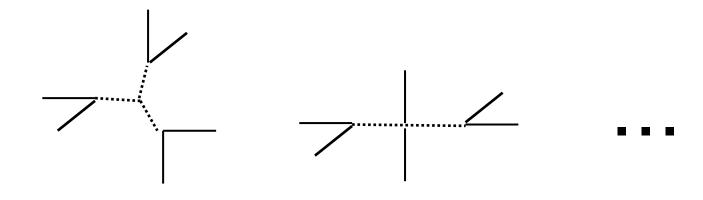
$$A_{ija} = W_{ia}^{(1)} W_{ja}^{(2)}, \quad D_{cmn} = W_{mc}^{(5)} W_{nc}^{(6)}$$

 $B_{akb} = W_{ak}^{(3)} \delta_{ab}, \quad C_{blc} = W_{lb}^{(4)} \delta_{bc}$



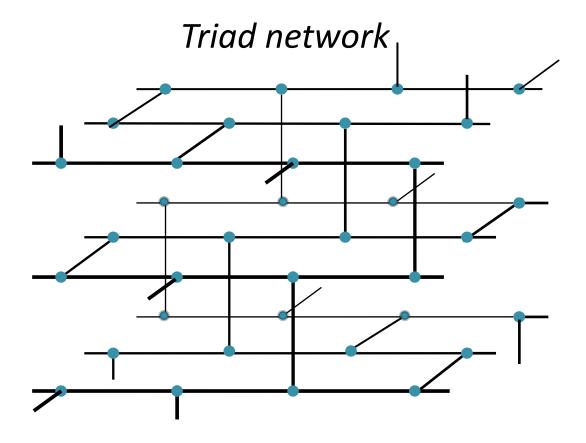
Many kinds of triad representation





tetrad mixture

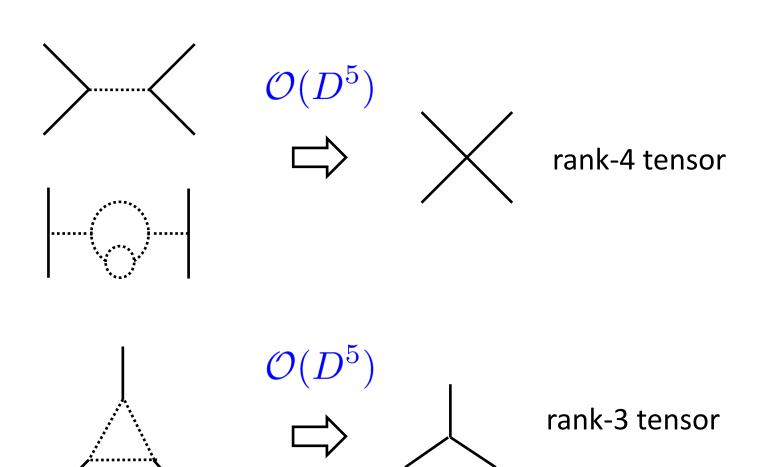
Triad networks and RGs



The cost of RGs on a triad network is naturally reduced because ...

The cost for rank-3 tensors

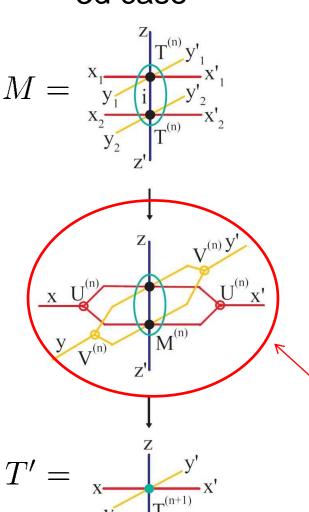
network of rank-3 tensors



Higher-order TRG (HOTRG)

3d case

Xie et al., 2012



$$M=T\cdot T$$

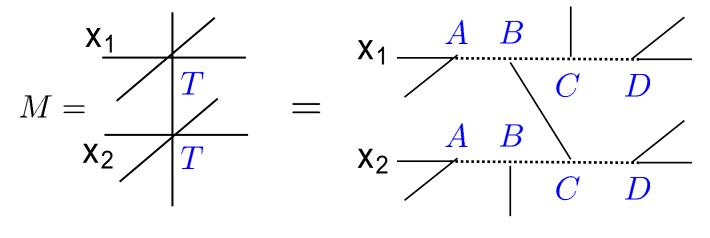
$$(UMM)U^\dagger)_{XX'}=\sigma_X\delta_{XX'}$$

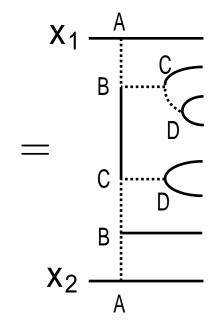
$$U_{X,x_1x_2}: ext{projector}$$
 $\mathcal{O}(D^{2d+2})$

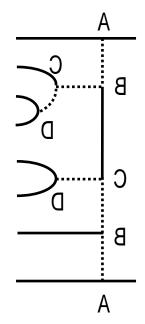
Contractions for making a renormalized tensor

$$\mathcal{O}(D^{4d-1})$$

M in the triad representation

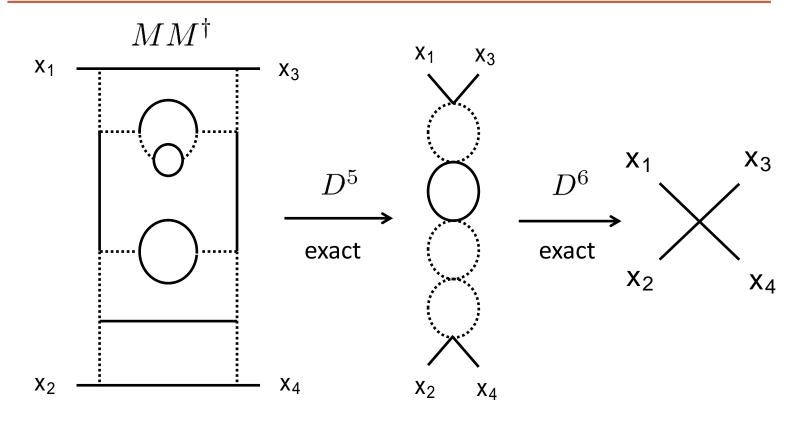






 M^{\dagger} is a mirror image of M.

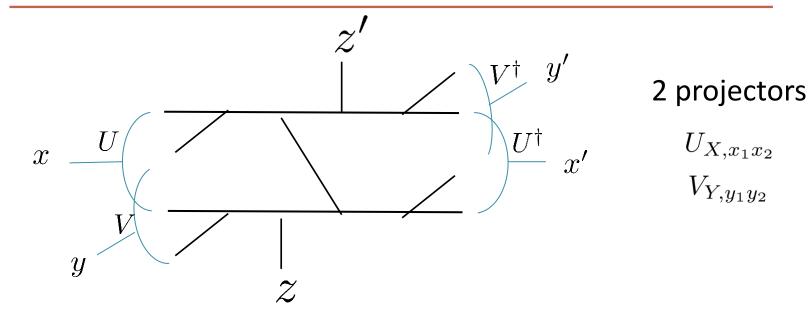
Steps of making projectors

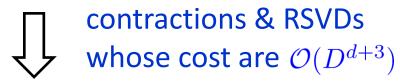


projectors U can be exactly prepared at an $\mathcal{O}(D^6)$ cost in any dimension!

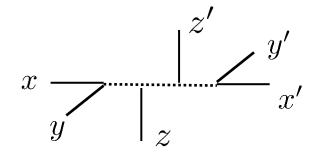
 $\mathcal{O}(D^5)$ by using a randomized SVD

Contraction of two triads

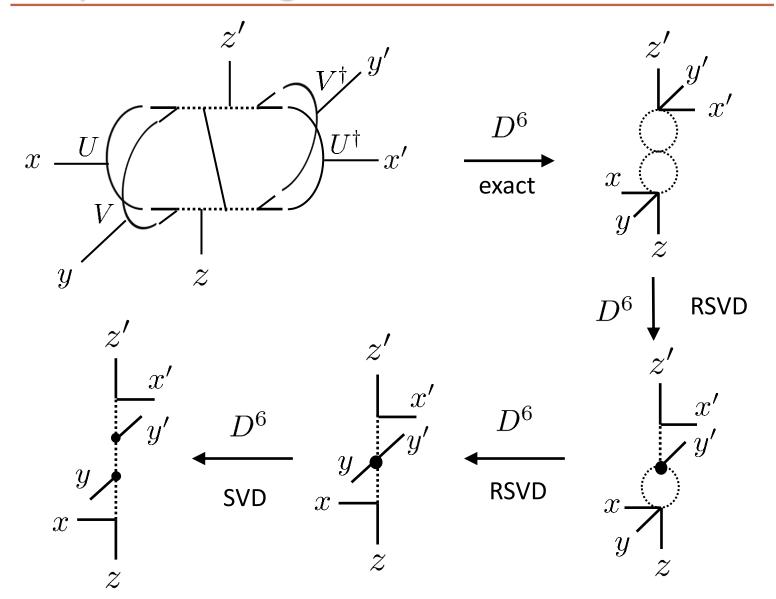




renormalized triad



Steps of making a renormalized triad



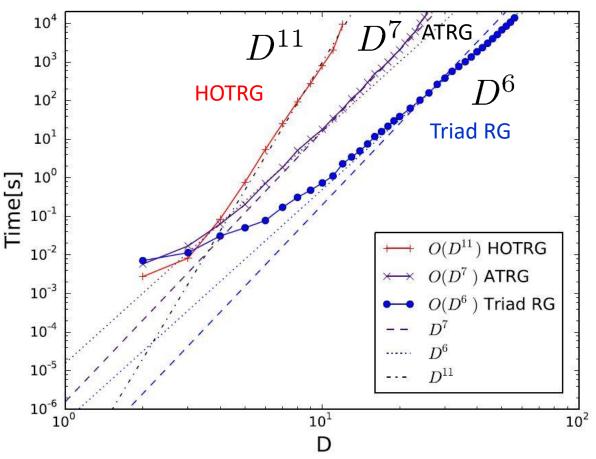
Theoretical cost

	•	1	
dim	α	nality	
CHILL	-11510	nality	
U	C		
		_	

	2	3	4	d
TRG	D^5			
HOTRG	D^7	D^{11}	D^{15}	D^{4d-1}
Anisotropic TRG [Adachi et al.,2019]	D^5	D^7	D^9	D^{2d+1}
Triad TRG (this work)	D^5	D^6	D^7	D^{d+3}

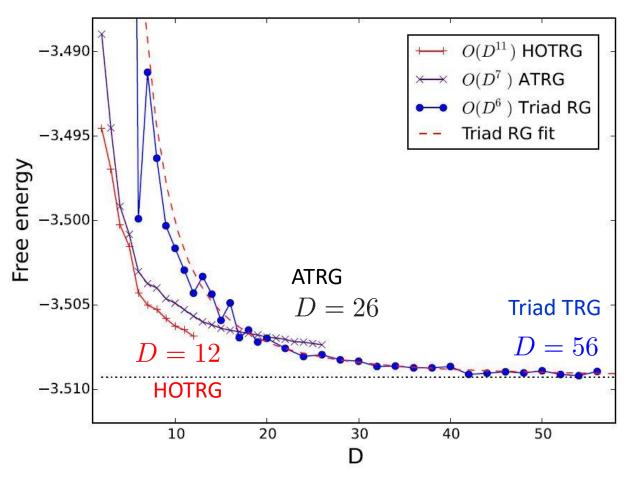
Numerical test in 3d Ising model at Tc

Computational time vs. D



Theoretical D-dependence is properly reproduced in actual computations.

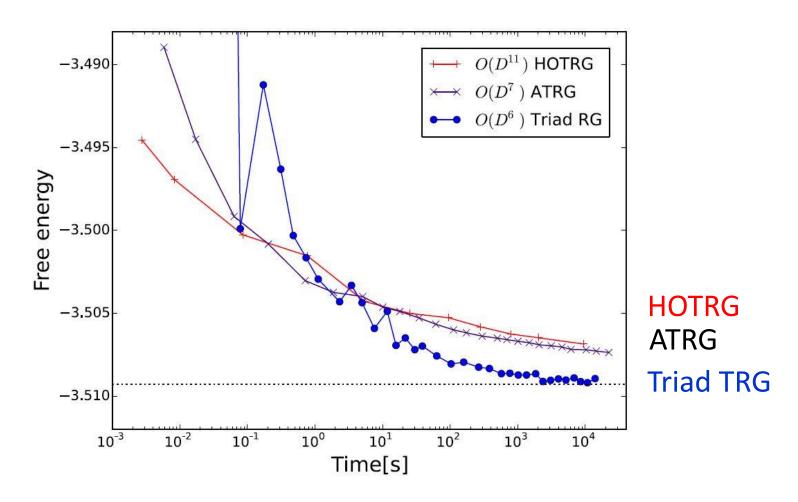
D-dependence of free energy



a few hours using laptop computers

The Triad TRG method shows good convergence as D increases.

Free energy vs. computational time



The other methods need much more time to approach a converged value around -3.509.

4. Summary and future outlook

Conclusions

The tensor network approach:

Good: no sign problem & no statistical error free energy is directly calculable. the large volume limit is easily taken.

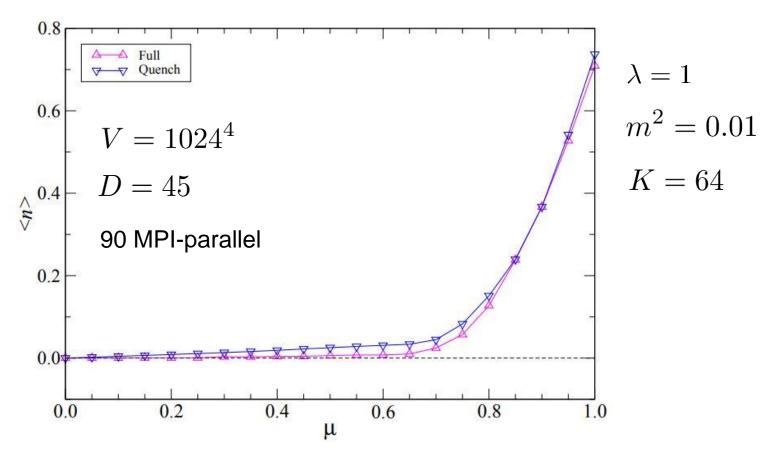
Bad: cost is high for $d \geq 3$ and large internal DoF.

 We find that the Gaussian quadrature works for scalar theory.

 The triad networks are effective in higher dimensions (at least, in 3 dimensions).

4d complex scalar theory at finite density

Akiyama, DK, Kuramashi, Yamashita, Yoshimura. JHEP 2020



The Silver braze is obtained for 4d complex scalar theory at finite density using the parallel computation of ATRG with a supercomputer.

future outlook

fermions

The Grassmann-TRG [Gu et al.,2010] is reformulated in [Akiyama-Kadoh,2020].

- lattice gauge theory
 character expansion of gauge group
- further improvements of triad TRG
 many kinds of triad → many variants of the RG
- high cost for large internal DoF (such as large N gauge theories)

quantum computer?