

# Anisotropic tensor renormalization group and BTRG

#### The University of Tokyo, Tsuyoshi Okubo

Ref. D. Adachi, T. Okubo, and S. Todo, arXiv:1906.02007

#### Collaborators

#### Department of Physics, The University of Tokyo



Daiki Adachi



Synge Todo

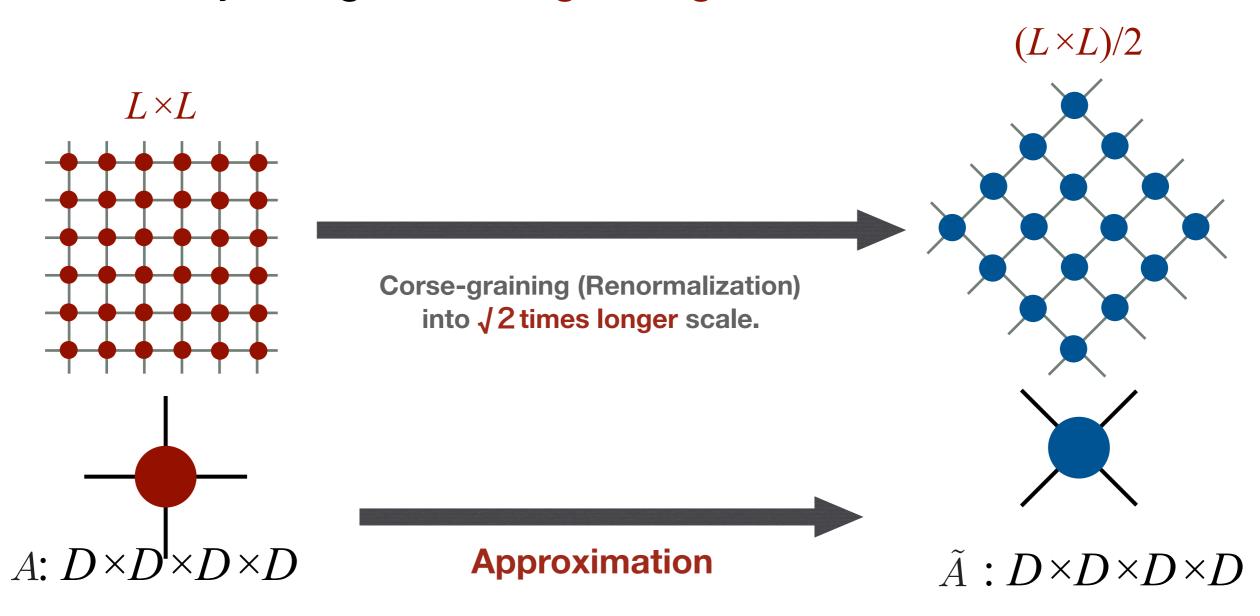


#### Contents

- Tensor renormalization group for high dimensions
- Anisotropic tensor renormalization group: ATRG
- Bond-weighted TRG: BTRG....
- Summary

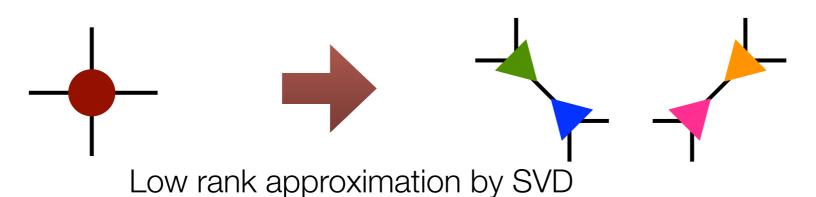
## Tensor network renormalization group

Purpose: approximate contraction of tensor network by using "coarse-graining" of the network



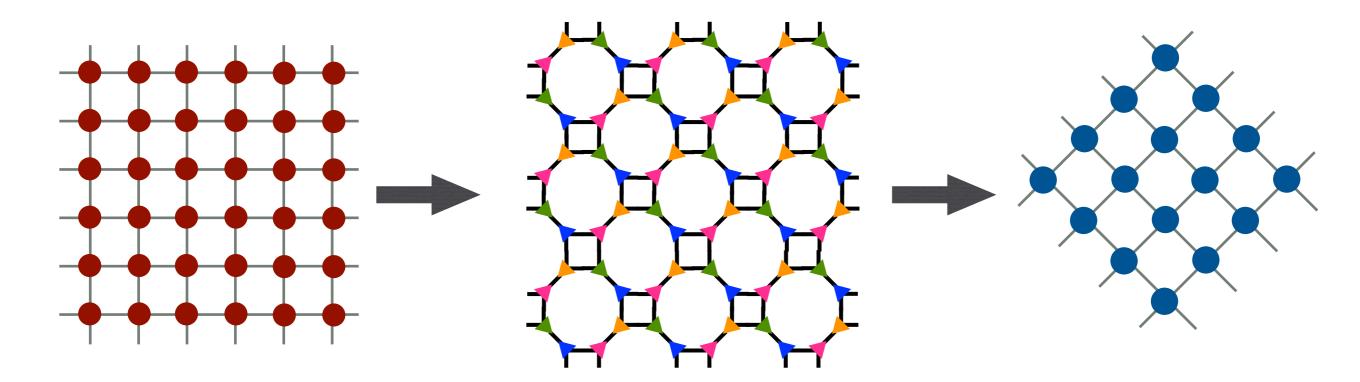
# Tensor Renormalization Group (TRG) algorithm

TRG M. Levin and C. P. Nave, Phys. Rev. Lett. **99**, 120601 (2007)



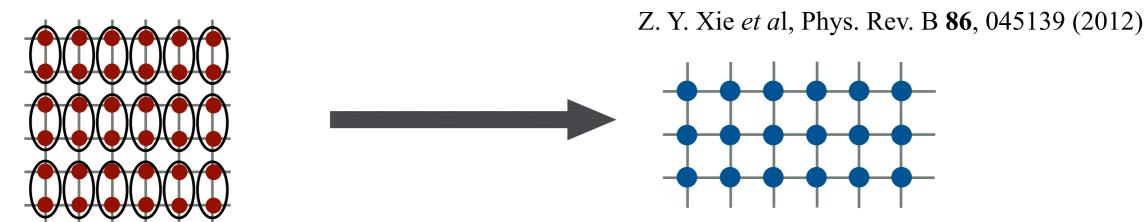
Computation cost:  $O(D^5)$ 

Memory:  $O(D^3)$ 



# Higher Oder Tensor Renormalization Group (HOTRG)

Anisotropic coarse-graining by using HOSVD instead of SVD



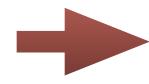
Better accuracy than TRG, although,

Computation cost:  $O(D^7) > O(D^5)$  (TRG)

## Application to high dimensions

#### Interests in

- 3d classical systems
- 2d and 3d quantum systems
- Much higher dimensions...



We want to perform tensor network RG for high dimensions!

However,

**TRG**: Not easy to generalize to high dimensions.

**HOTRG**: Easy to generalize to high dimensions, but its cost is  $O(D^{4d-1})$ 

Is it possible to construct lower cost algorithm?

 $d=3: O(D^{11})$ 

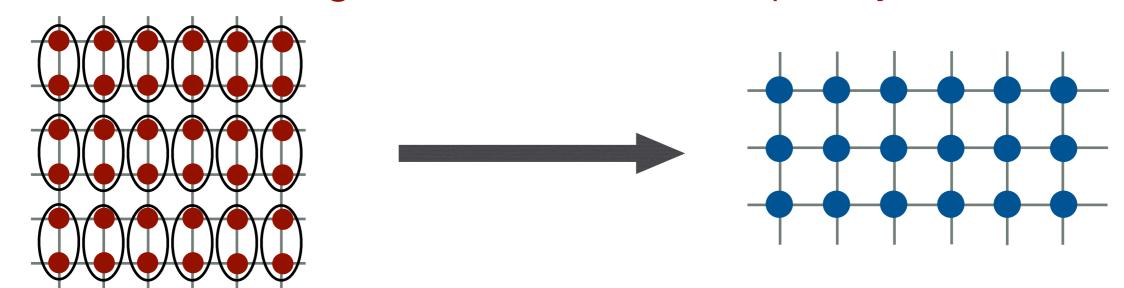
 $d=4: O(D^{15})$ 

# Anisotropic TRG = ATRG

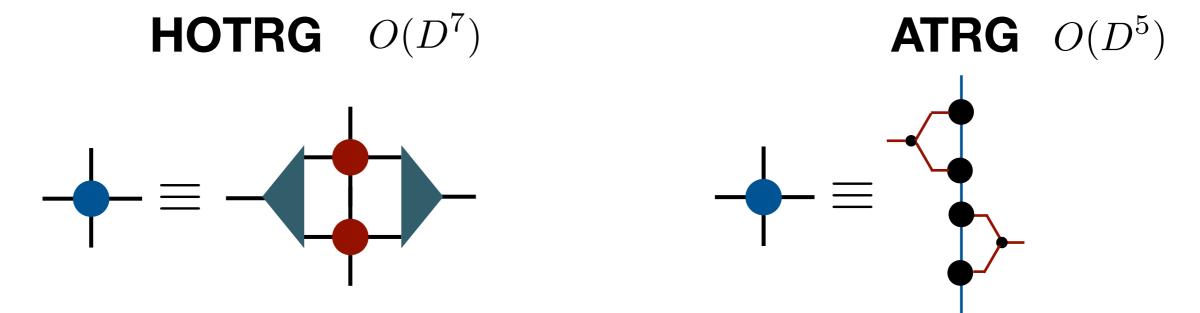
D. Adachi, T. Okubo, and S. Todo, arXiv:1906.02007

## Central idea of Anisotropic TRG

In ATRG, we coarse-grain tensors anisotropically as HOTRG:

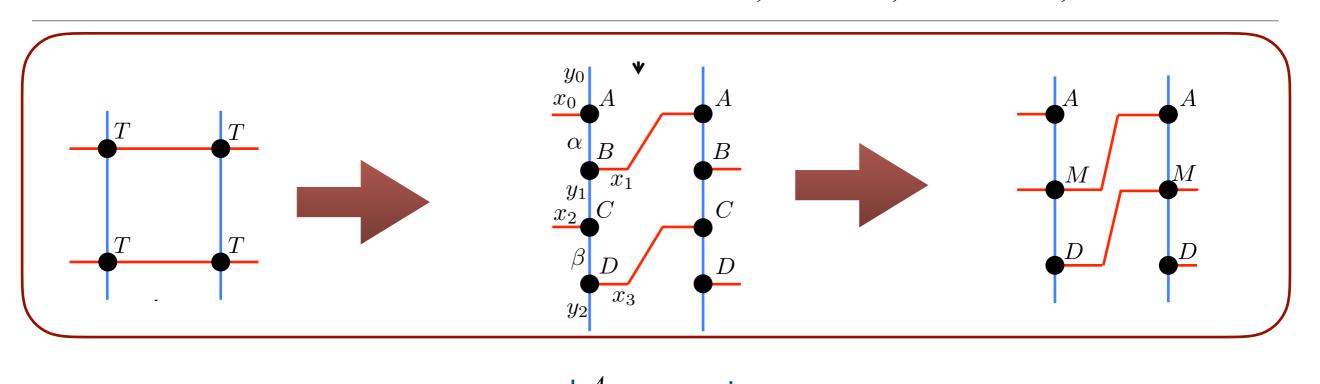


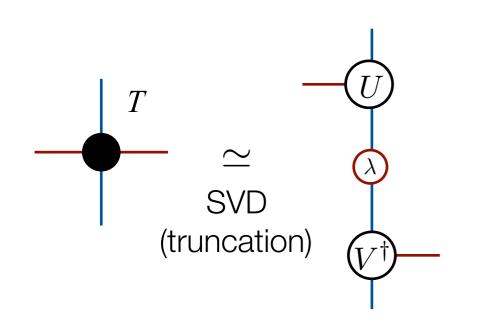
In order to reduce the computation cost, we decompose the local tensor into small pieces before performing coarse-graining.

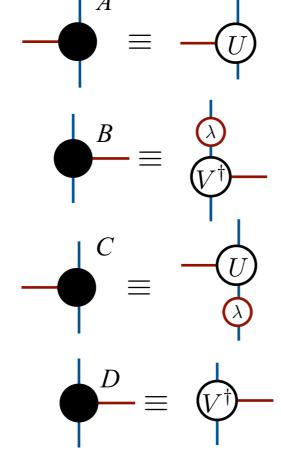


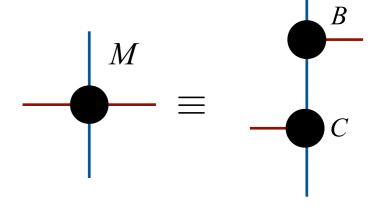
# Recipe of ATRG

D. Adachi, T. Okubo, and S. Todo, arXiv:1906.02007





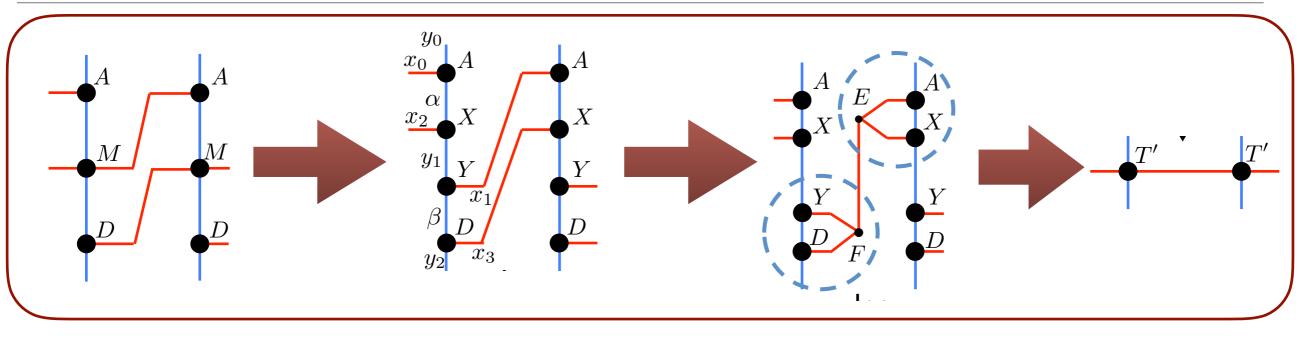


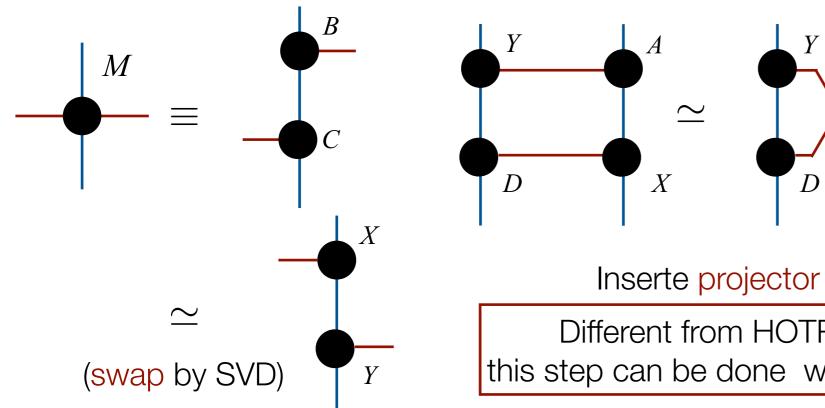


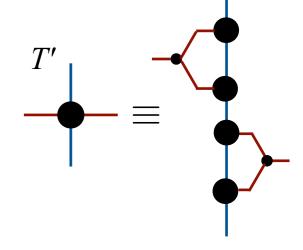
# Recipe of ATRG

D. Adachi, T. Okubo, and S. Todo, arXiv:1906.02007

X







Different from HOTRG, this step can be done with  $O(D^5)$ 

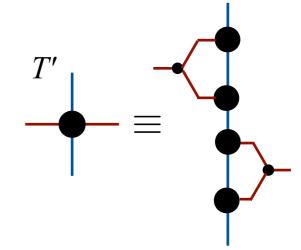
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# Summary of 2D ATRG

D. Adachi, T. Okubo, and S. Todo, arXiv:1906.02007

## Memory storage: $O(D^3)$

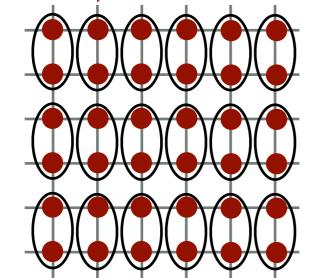
\* We do not explicitly create 4-leg tensor. (We need only 3-leg tensors!)

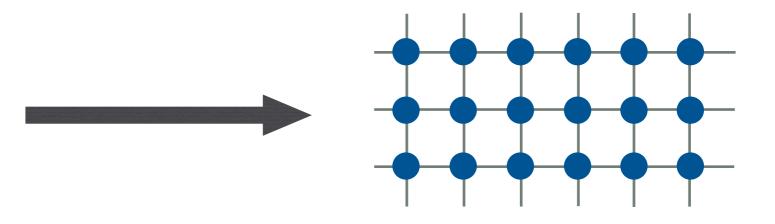


#### Computation time: $O(D^5)$

\* By using *partial SVD* technique, such as the Arnoldi method, we can reduce SVD cost.

We can perform HOTRG like anisotropic coarse-graining with smaller cost!





## Generalization to high dimension

D. Adachi, T. Okubo, and S. Todo, arXiv:1906.02007

We can easily generalize this to high dimensions

#### **Key points:**

Before coarse graining, we decompose a local tensor into two tensors.



\*We may consider more decompositions.

Memory storage:  $O(D^{d+1})$ 

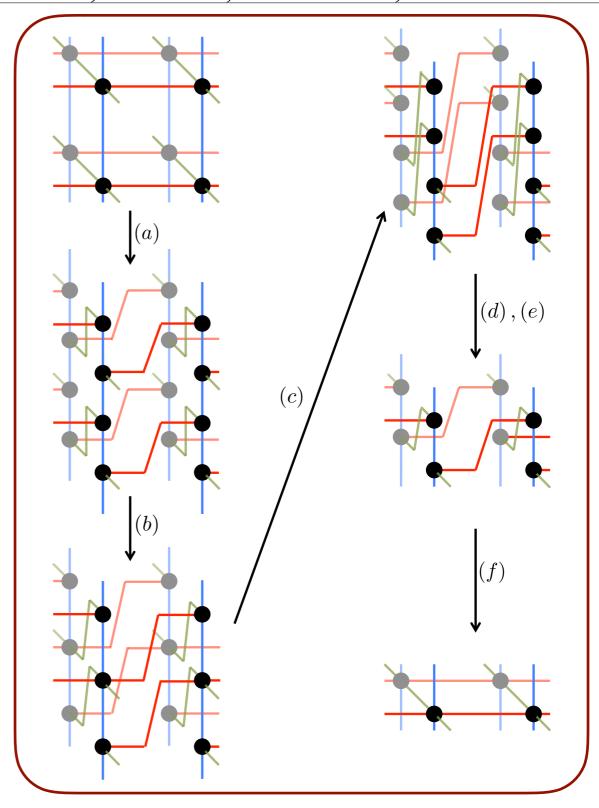
Computation time:  $O(D^{2d+1})$ 



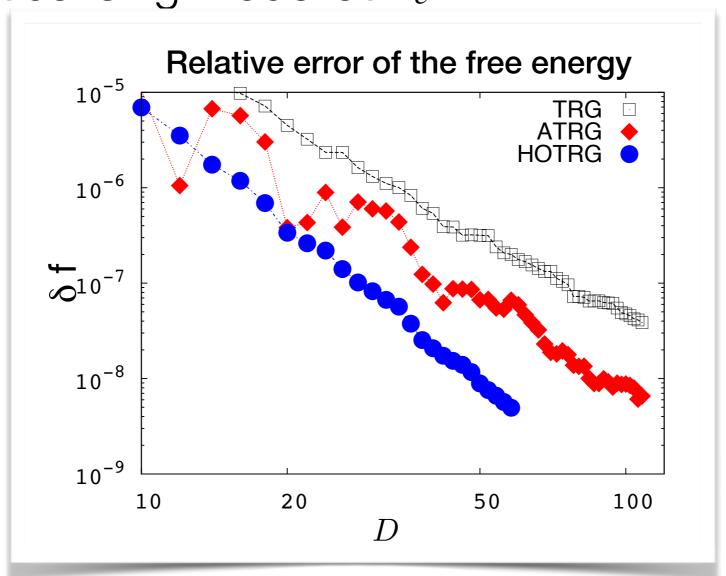
#### HOTRG

Memory:  $O(D^{2d})$ 

Time:  $O(D^{4d-1})$ 



#### 2d square lattice Ising model at $T_c$



At the same bond dimension, accuracy of ATRG is between those of TRG and HOTRG.

**HOTRG** is better?



No so straightforward!

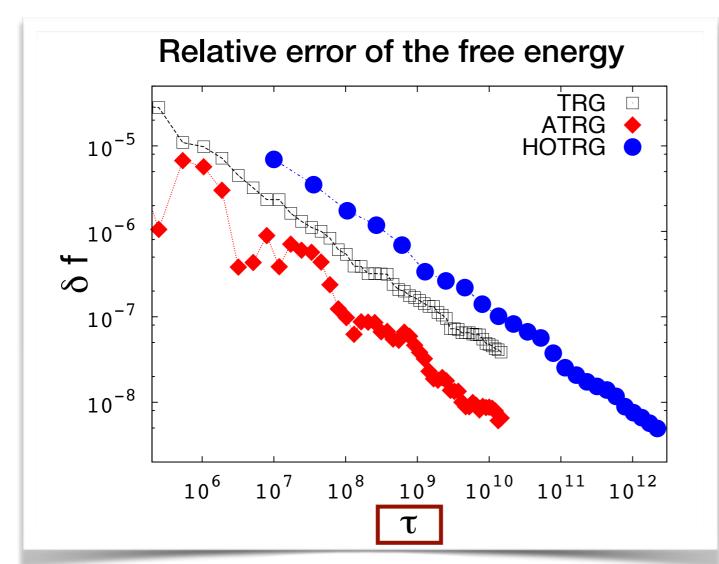
#### Benchmarks

The computation costs are different between ATRG and HOTRG.

TRG, ATRG:  $O(D^5)$ HOTRG:  $O(D^7)$ 

Leading order computation time:

$$\tau \equiv \begin{cases} D^5 & \text{TRG and ATRG} \\ D^7 & \text{HOTRG} \end{cases}$$



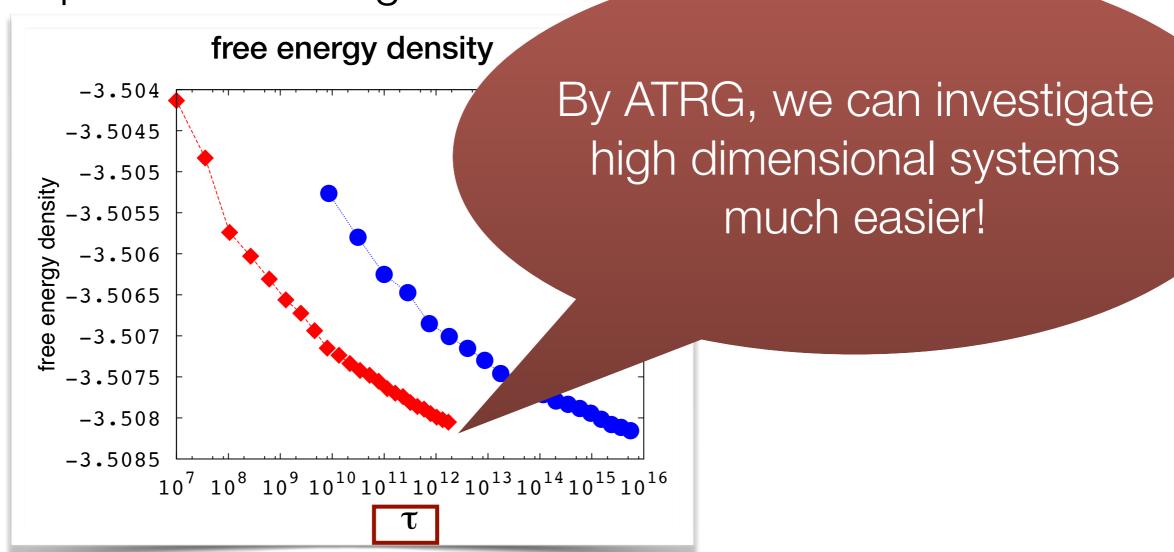
#### ATRG is the best!

(ATRG is ~10<sup>2</sup> faster than HOTRG!)

#### Benchmarks

D. Adachi, T. Okubo, and S. Todo, arXiv:1906.02007

**3d** square lattice Ising model at *T* 



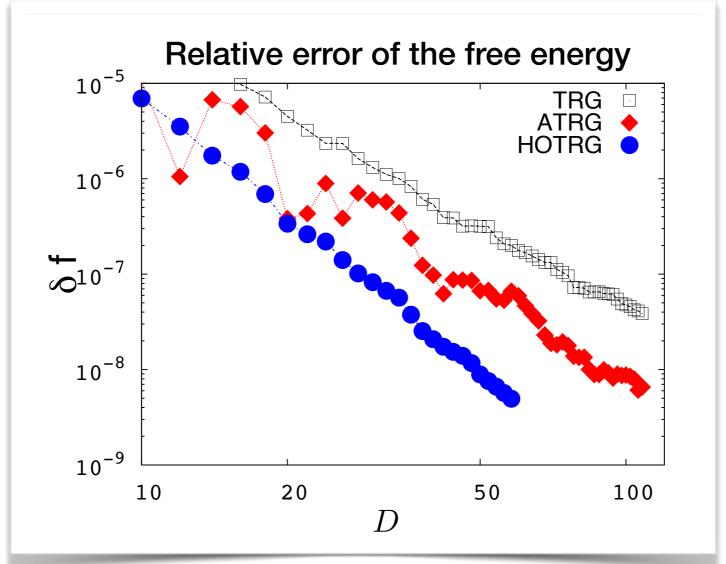
In order to obtain same free energy value, ATRG needs ~10<sup>3</sup> times smaller computation cost than HOTRG.

# Bond-weighted TRG = BTRG

D. Adachi, T. Okubo, and S. Todo, in preparation

# Why ATRG is better than TRG?

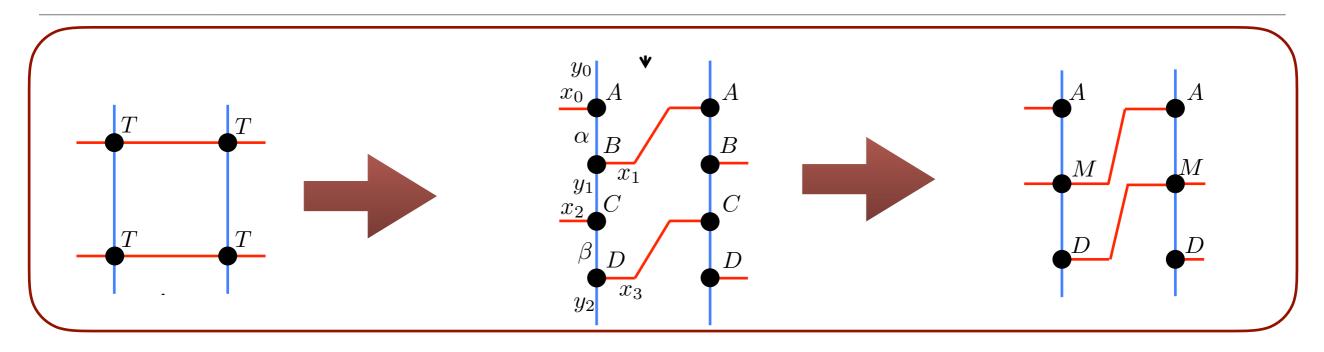
2d square lattice Ising model at  $T_c$ 

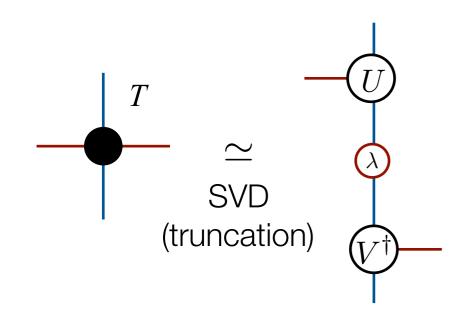


ATRG is better than TRG, although the cost is same.

Why?

# Key point: weight of bonds



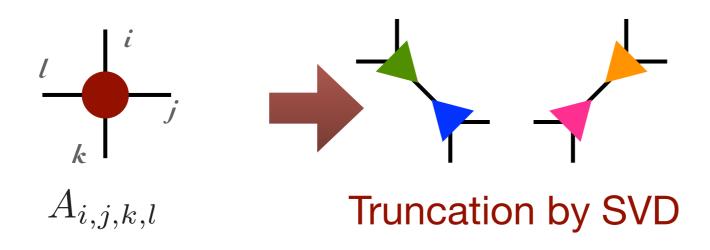


In ATRG, we assign singular values to only one tensor.



In TRG, usually it is divided to both tensors.

## Disadvantage of local SVD truncation



Local SVD is not necessarily the best "D-rank" approximation for the whole TN (the partition functions).

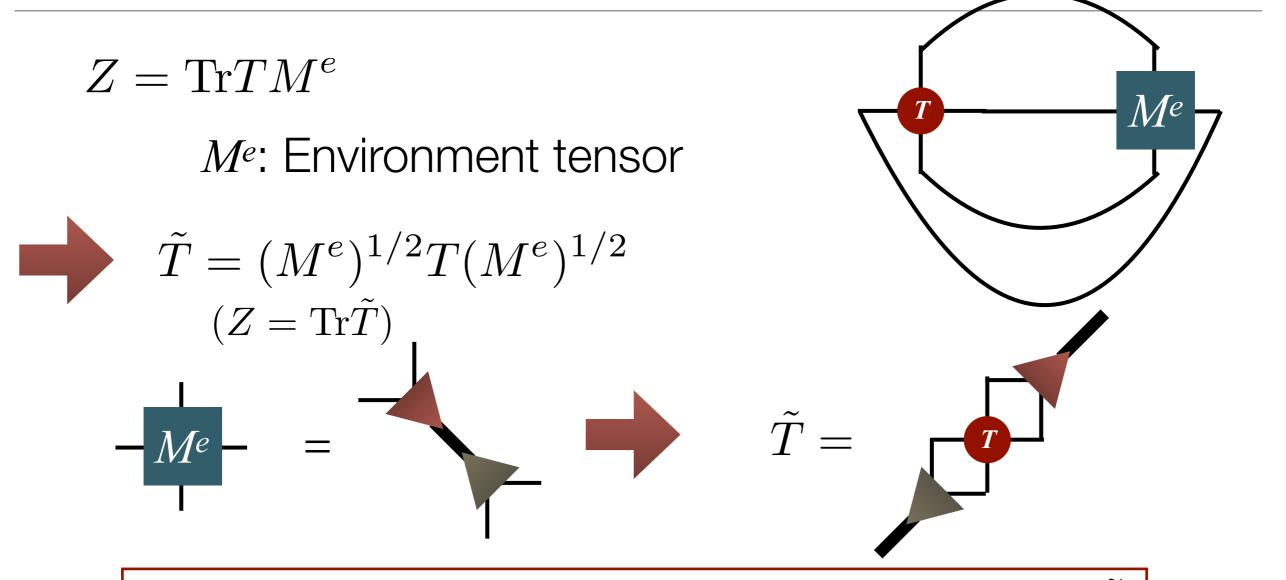


We need to consider the "environment" for better truncations.

This is the key idea of SRG (second renormalization)

## Basic strategy of SRG

Z. Y. Xie, et al Phys. Rev. Lett. 103, 160601 (2008).



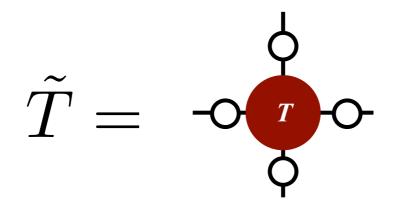
Better truncation: truncation based on SVD of  $\tilde{T}$ 

#### **Disadvantage of SRG:**

In order to obtain  $M^e$  we need iterative calculations, which increases computation cost.

#### Mean field environment

As an approximated  $M^e$ , we might use "weight" of bonds as mean-field environment.



—O—: Diagonal non-negative matrix

How can we estimate the bond weight?

- We can estimate it from the singular values of a local tensor.
  - 1. The weights are repeatedly improved by the singular values of  $ilde{T}$ .

Mean-field SRG. H.-H. Zhao et al, PRB 81, 174411 (2010).

2. Alternatively, we can consider the renormalization of extended tensor networks which contains bond weight explicitly.



# Summary

- We introduced Anisotropic TRG (ATRG)
  - By using ATRG, we can contract a d-dimensional TN with O(D<sup>2d+1</sup>) computation cost and O(D<sup>d+1</sup>) memory.
  - When we look accuracy per computation time, ATRG is much more efficient than HOTRG.
- We considered TRG of bond-weighed TN (BTRG)
  - By choosing proper exponent, BTRG shows better accuracy than HOTRG.
  - BTRG is easily generalized for other lattices and HOTRG.