

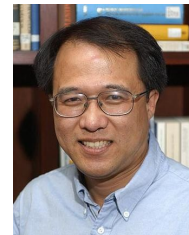
Loop optimization for tensor network renormalization

Shuo Yang

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Zheng-Cheng Gu
(CUHK, PI)

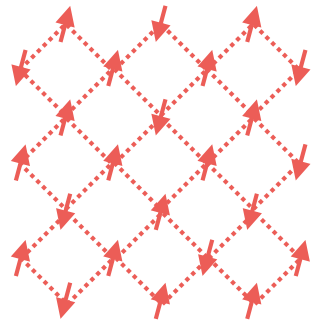


Xiao-Gang Wen
(MIT, PI)

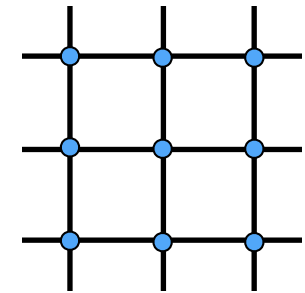
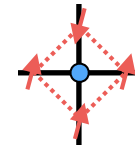
Overview

tensor network + renormalization group = tensor network renormalization

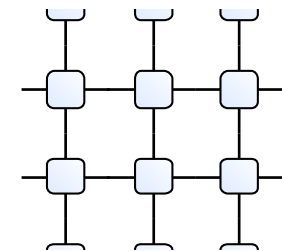
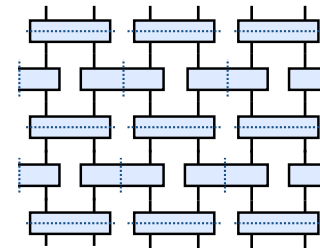
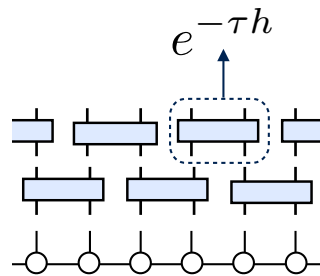
Partition function of classical statistical system



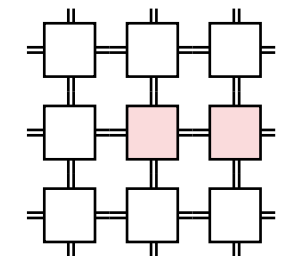
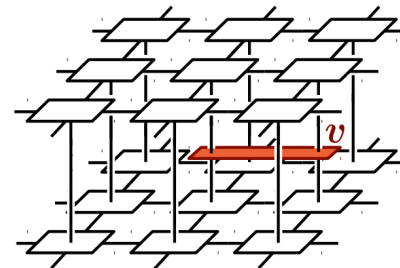
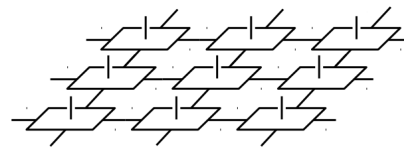
$$Z = \sum_{\{\sigma\}} \exp(\beta \sum_{\langle ij \rangle} \sigma_i \sigma_j)$$



Euclidean path integral of 1D quantum system



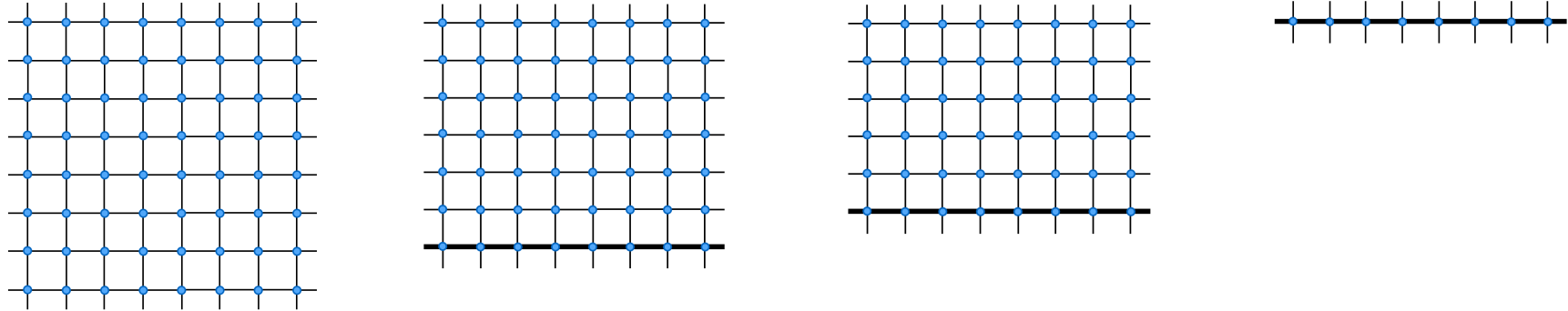
Physical properties of 2D quantum system



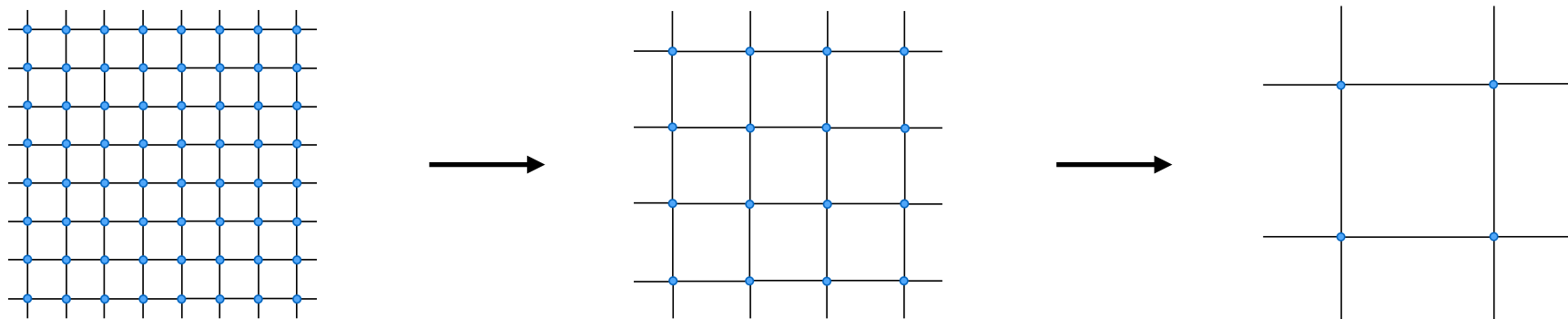
Overview

tensor network + renormalization group = tensor network renormalization

■ boundary MPS



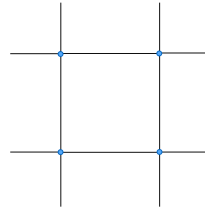
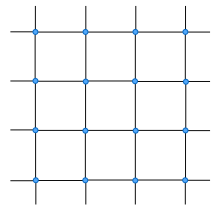
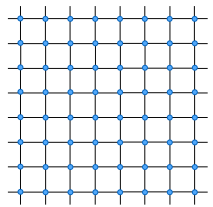
■ fully isotropic coarse-graining



Overview

tensor network + renormalization group = tensor network renormalization

How the physics change with scale



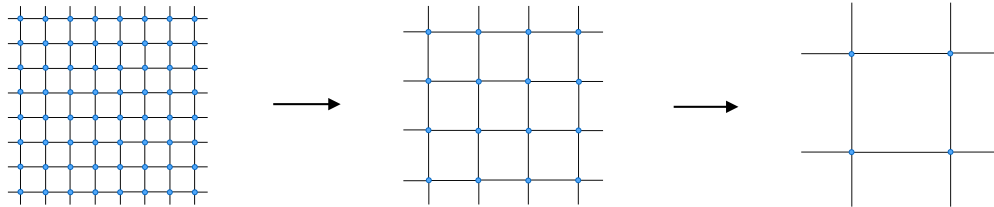
scale transformation

- Non-perturbative approach, suitable for strongly correlated systems
- Reproduce long-range physics

Overview

tensor network + renormalization group = tensor network renormalization

How the physics change with scale



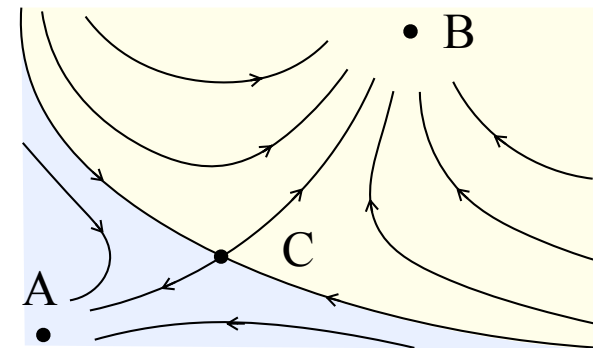
scale transformation

- Non-perturbative approach, suitable for strongly correlated systems
- Reproduce long-range physics

Aim

- Remove short-range entanglement / correlations
- Generate proper RG flow & correct fixed points
- Recover scale invariance at criticality

RG flow

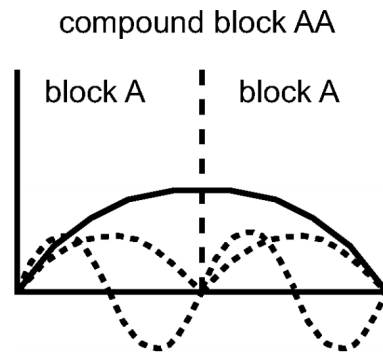


Real space renormalization group

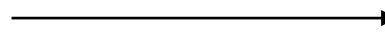
1D quantum:

Real space analogue of NRG
Wilson (1975)
breakdown for “particle in a box” problem

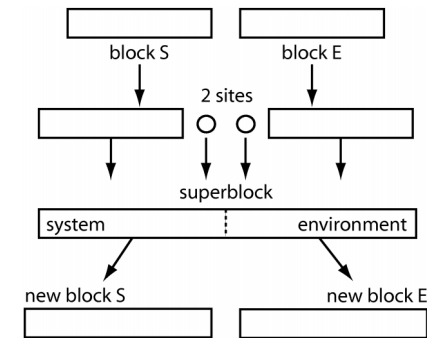
real space NRG



entanglement

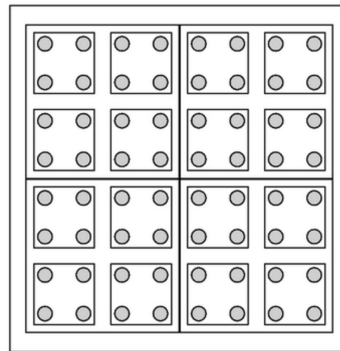


DMRG S. White (1992)



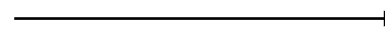
2D classical:

block spin

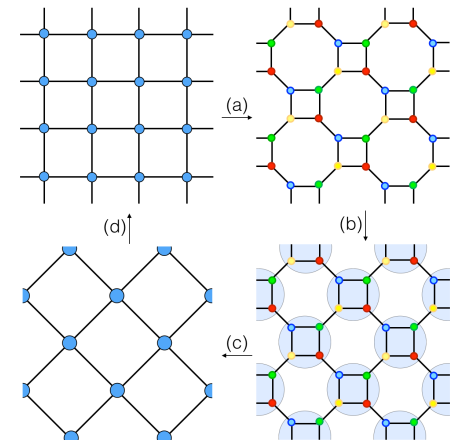


L.P. Kadanoff (1966)

entanglement



TRG Levin & Nave (2007)

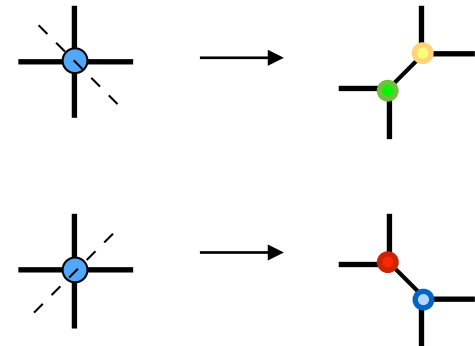


Tensor renormalization group

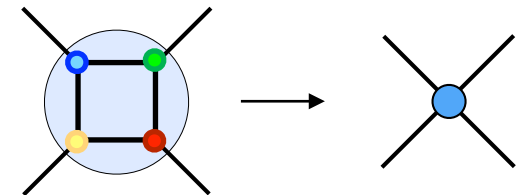
Levin & Nave (2007) LN-TNR

Three steps

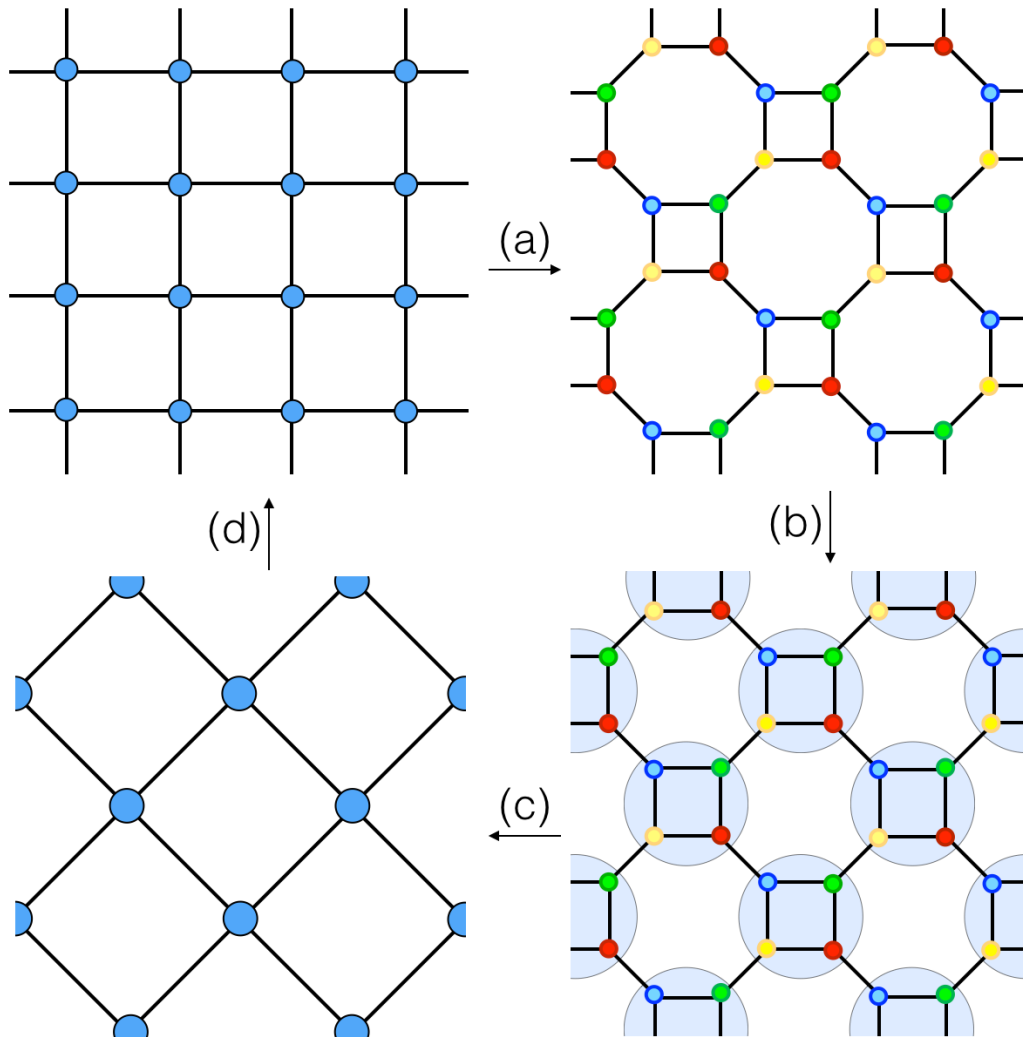
1. Deform tensors, make a truncation



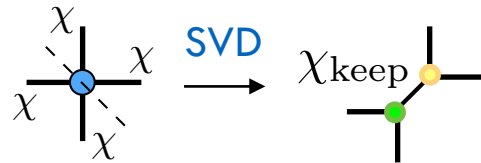
2. Coarse graining



3. Renormalize tensors
(multiply tensor by a constant factor)



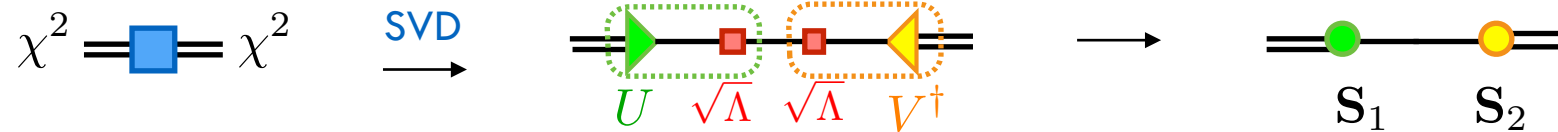
Tensor renormalization group



singular value decomposition (SVD)
only keep the largest χ_{keep} singular values

$$\left\| \left\| \begin{array}{c} \chi \\ \chi \\ \chi \\ \chi \end{array} \right\| - \chi_{\text{keep}} \begin{array}{c} |2 \\ |1 \end{array} \right\|^2$$

cost function
 $\|\mathbf{T} - \mathbf{S}_1 \cdot \mathbf{S}_2\|^2$



original

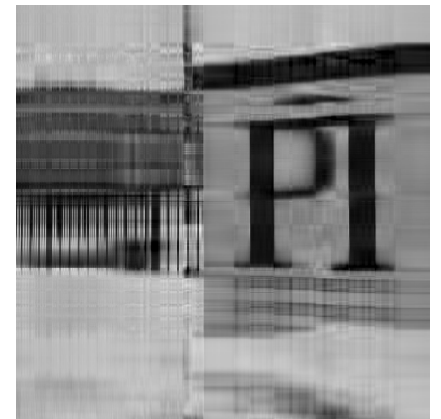
$$\chi_{\text{keep}} = \chi^2$$



$$\chi_{\text{keep}} = 2\chi$$



$$\chi_{\text{keep}} = \chi$$

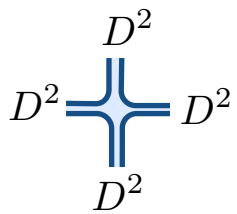
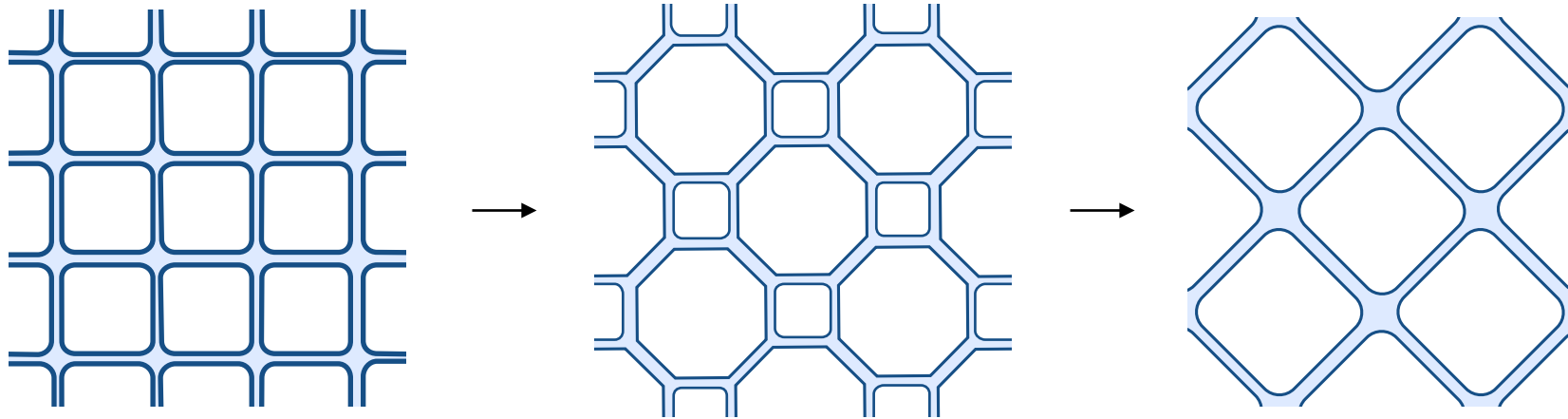


$$\chi_{\text{keep}} = \chi/4$$

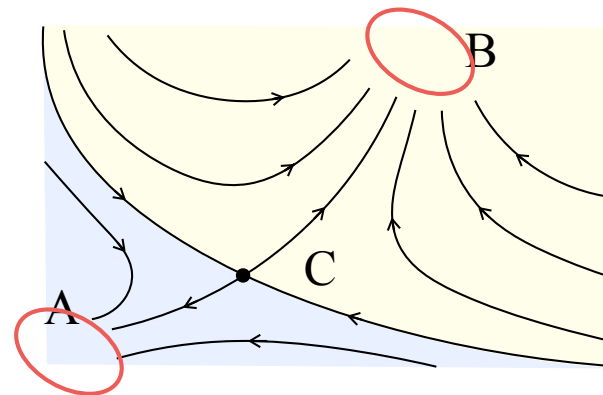
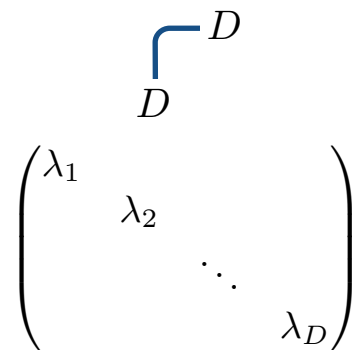
Drawbacks of LN-TNR

Off criticality

- cannot remove conner-double-line (CDL) tensors
- cannot give the correct structures of fixed points



$$\chi = D^2$$



Drawbacks of LN-TNR

At criticality

cannot explicitly recover scale invariance

cannot completely remove short-range entanglement

Drawbacks of LN-TNR

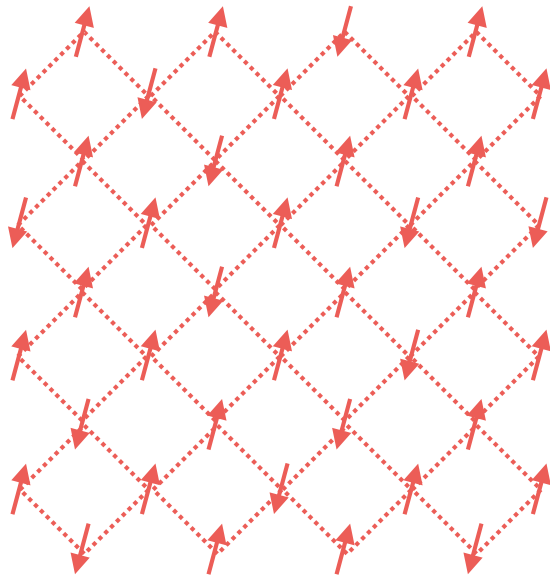
At criticality

cannot explicitly recover scale invariance

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Example

classical Ising model



partition function

$$Z = \sum_{\{\sigma\}} \exp(\beta \sum_{\langle ij \rangle} \sigma_i \sigma_j)$$

Drawbacks of LN-TNR

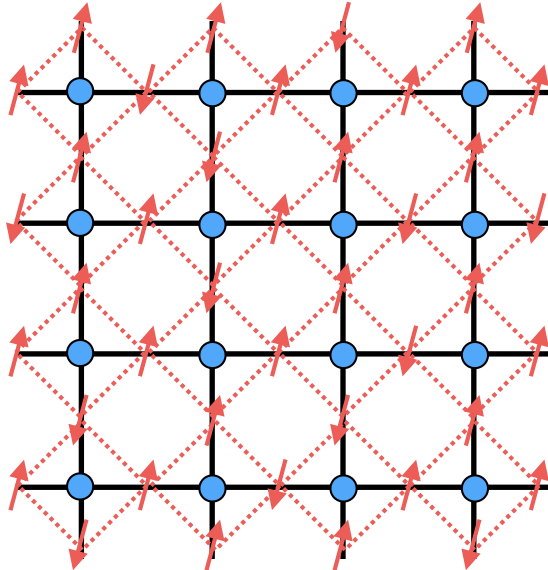
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Example

classical Ising model



partition function

$$Z = \sum_{\{\sigma\}} \exp(\beta \sum_{\langle ij \rangle} \sigma_i \sigma_j)$$

local tensor

$$\mathbf{T} = T_{u,l,d,r}^{\text{Ising}}$$

$$T_{1,2,1,2}^{\text{Ising}} = e^{-4\beta}, \quad T_{2,1,2,1}^{\text{Ising}} = e^{-4\beta},$$

$$T_{1,1,1,1}^{\text{Ising}} = e^{4\beta}, \quad T_{2,2,2,2}^{\text{Ising}} = e^{4\beta},$$

others = 1.

Drawbacks of LN-TNR

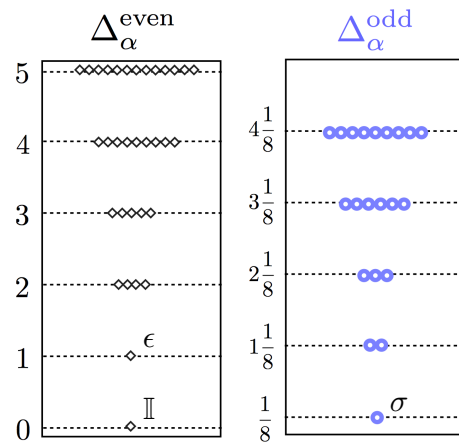
Ising CFT

central charge

$$c = 1/2$$

scaling

dimensions



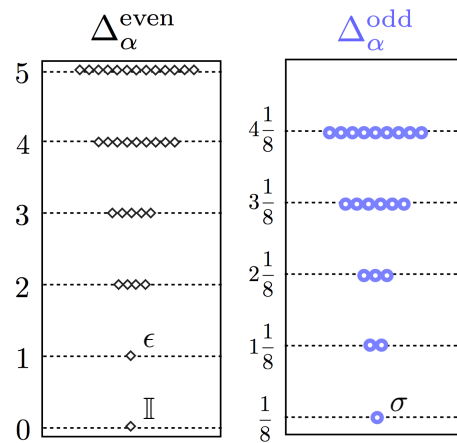
Drawbacks of LN-TNR

Ising CFT

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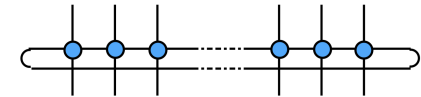
$$c = 1/2$$

scaling dimensions



Calculate scaling dimensions

transfer matrix



eigenvalues of the transfer matrix $\rightarrow c, \Delta_\alpha$

Zheng-Cheng Gu and Xiao-Gang Wen,
Phys. Rev. B **80**, 155131 (2009).

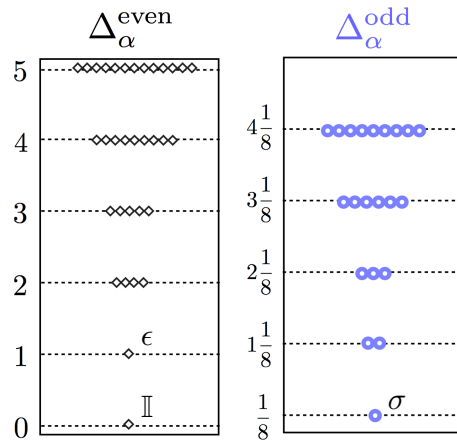
Drawbacks of LN-TNR

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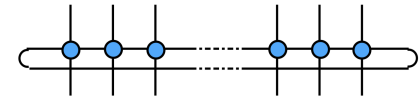
$$c = 1/2$$

scaling dimensions \rightarrow



Calculate scaling dimensions

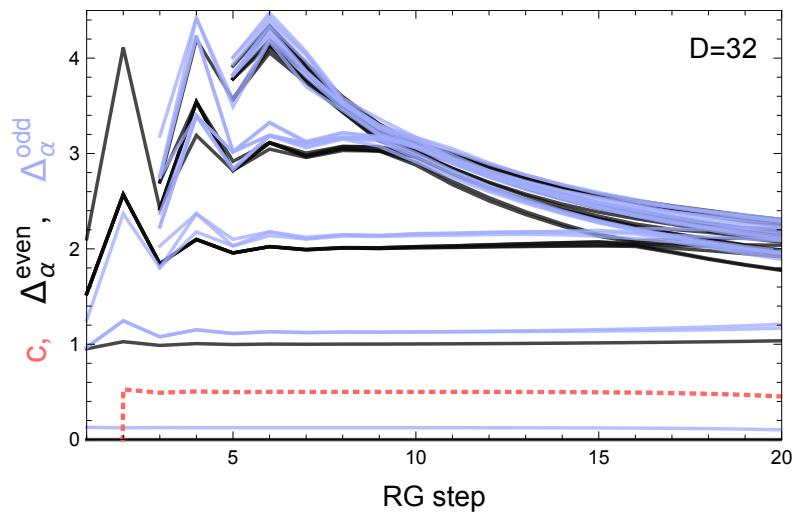
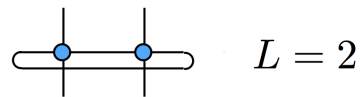
transfer matrix



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LN-TNR



scaling dimensions change with RG step
cannot explicitly recover scale invariance

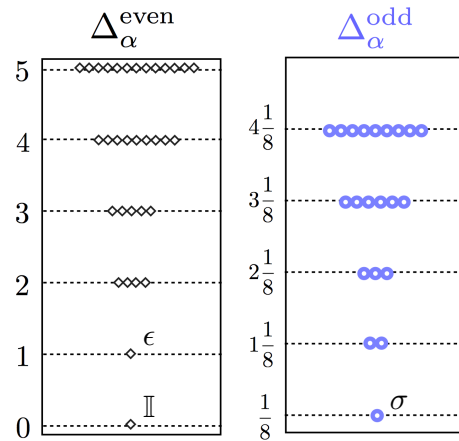
Drawbacks of LN-TNR

Ising CFT

central charge

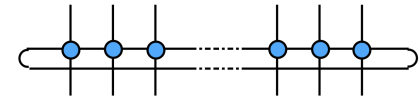
$$c = 1/2$$

scaling dimensions



Calculate scaling dimensions

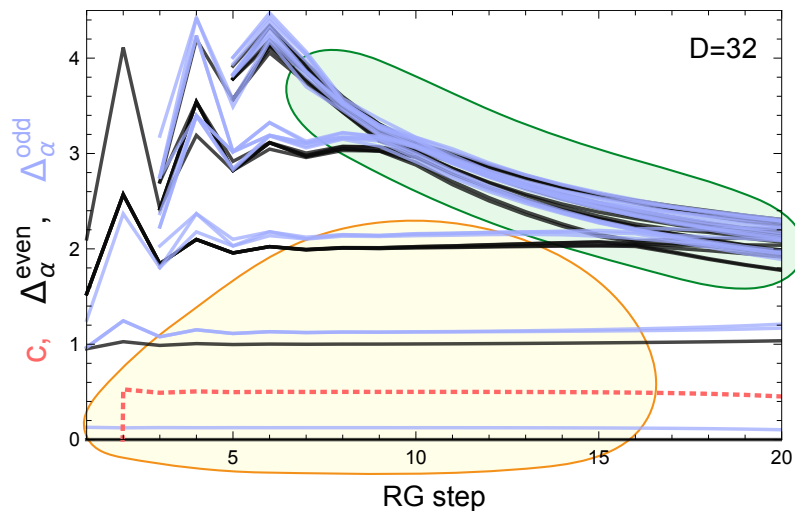
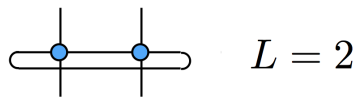
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Phys. Rev. B **80**, 151131 (2009).









LN-TNR

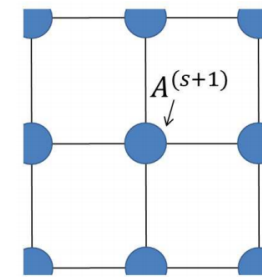
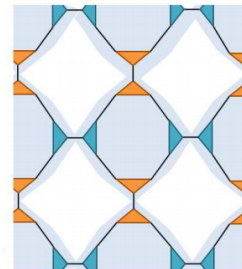
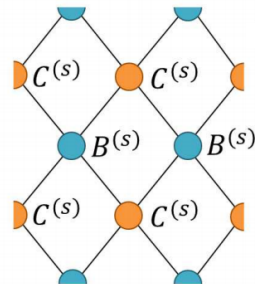
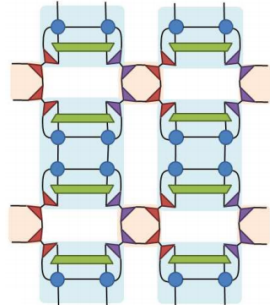
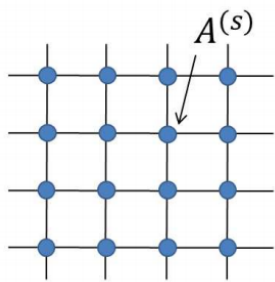


scaling dimensions change with RG step
cannot explicitly recover scale invariance











high-index parts will destroy low-index parts
accuracy is fine, stability is bad

Improvements of TRG

	off-critical	critical
TRG (Levin & Nave, 2006)		
SRG (Xie, Jiang, Weng, Xiang, 2008)		
TEFR (Gu & Wen, 2009)		
TNR (Evenbly & Vidal, 2014)		



Improvements of TRG

	off-critical	critical
TRG (Levin & Nave, 2006)		
SRG (Xie, Jiang, Weng, Xiang, 2008)		
TEFR (Gu & Wen, 2009)		
TNR (Evenbly & Vidal, 2014)		
Loop-TRG (current)		

Renormalization Group

Momentum space

- Tree level approximation — no loop

- Beyond tree level — one loop

Real space

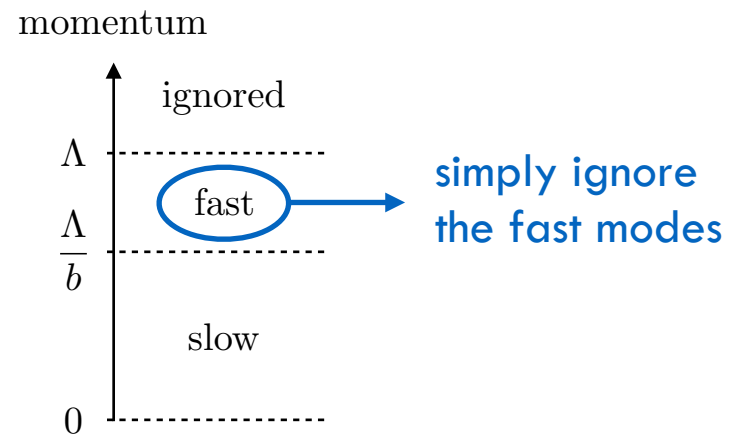
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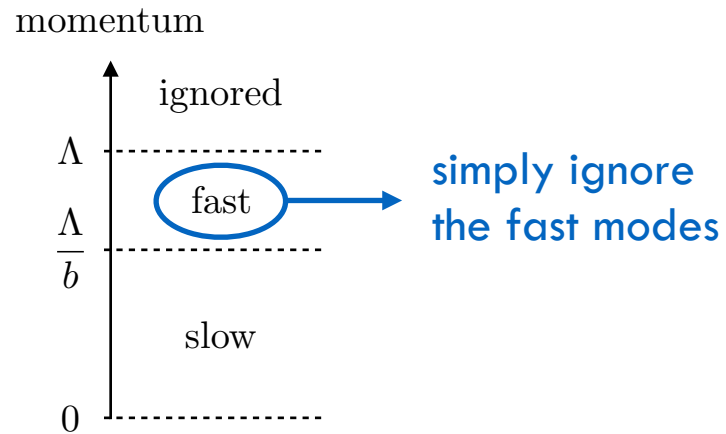
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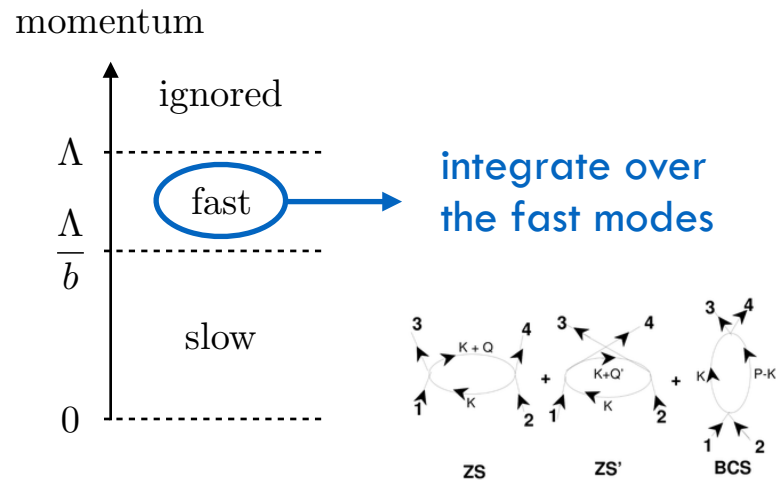
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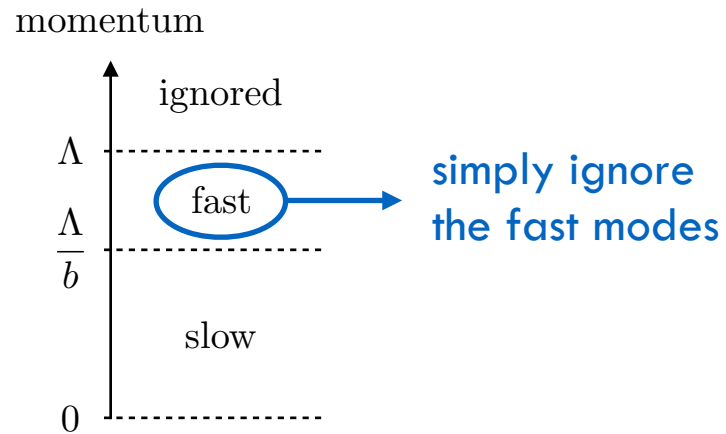
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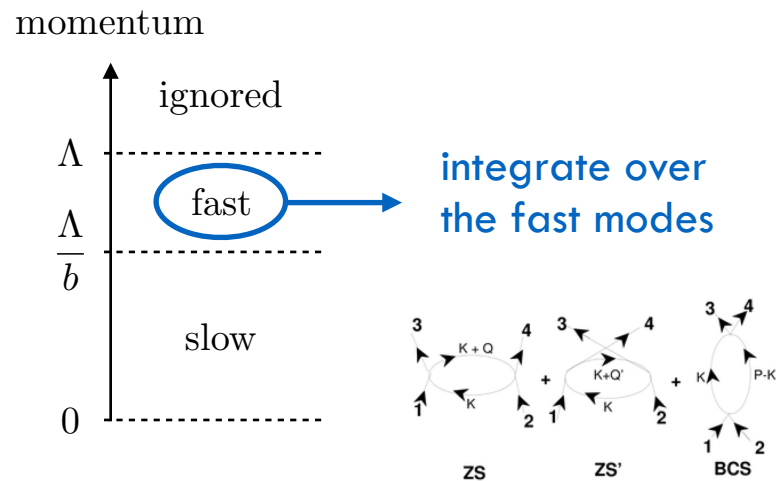
Renormalization Group

Momentum space

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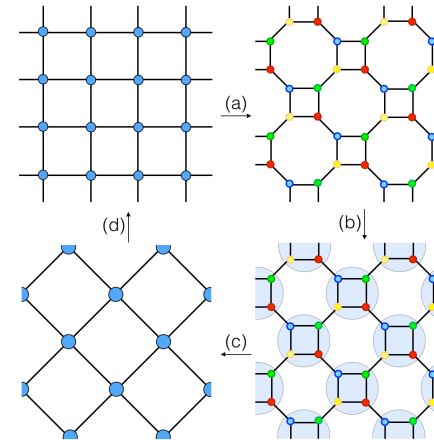


- Beyond tree level — one loop



Real space

- Tree level approximation — no loop



LN-TNR

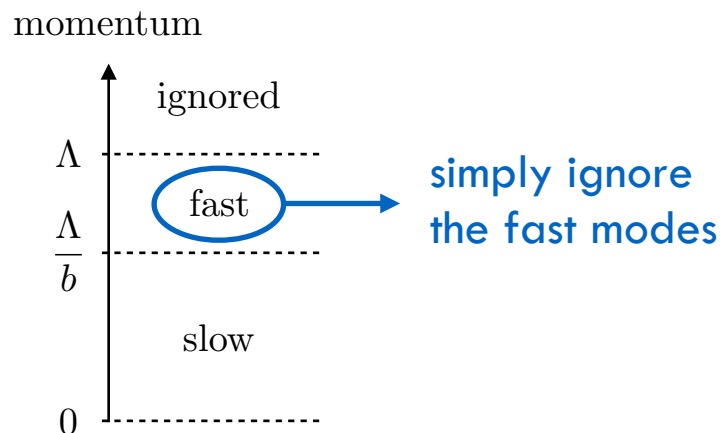
remove short-range entanglement by a local SVD

- Beyond tree level — one loop

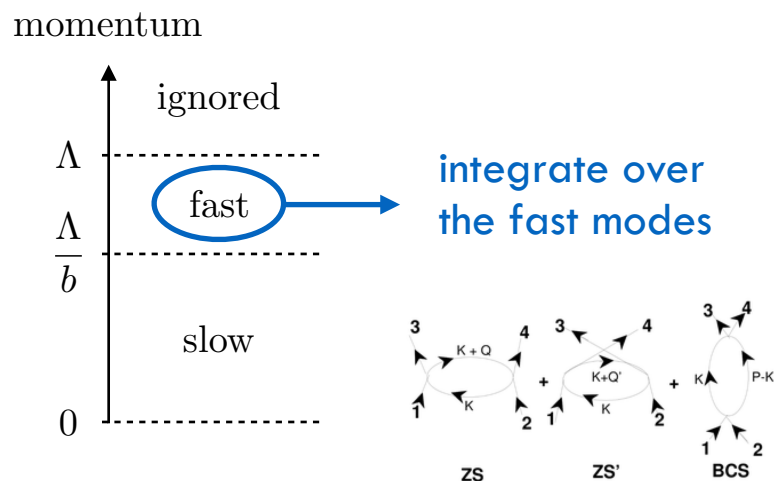
Renormalization Group

Momentum space

- Tree level approximation — no loop

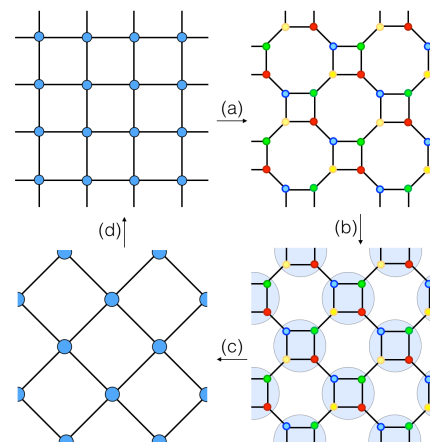


- Beyond tree level — one loop



Real space

- Tree level approximation — no loop



LN-TNR

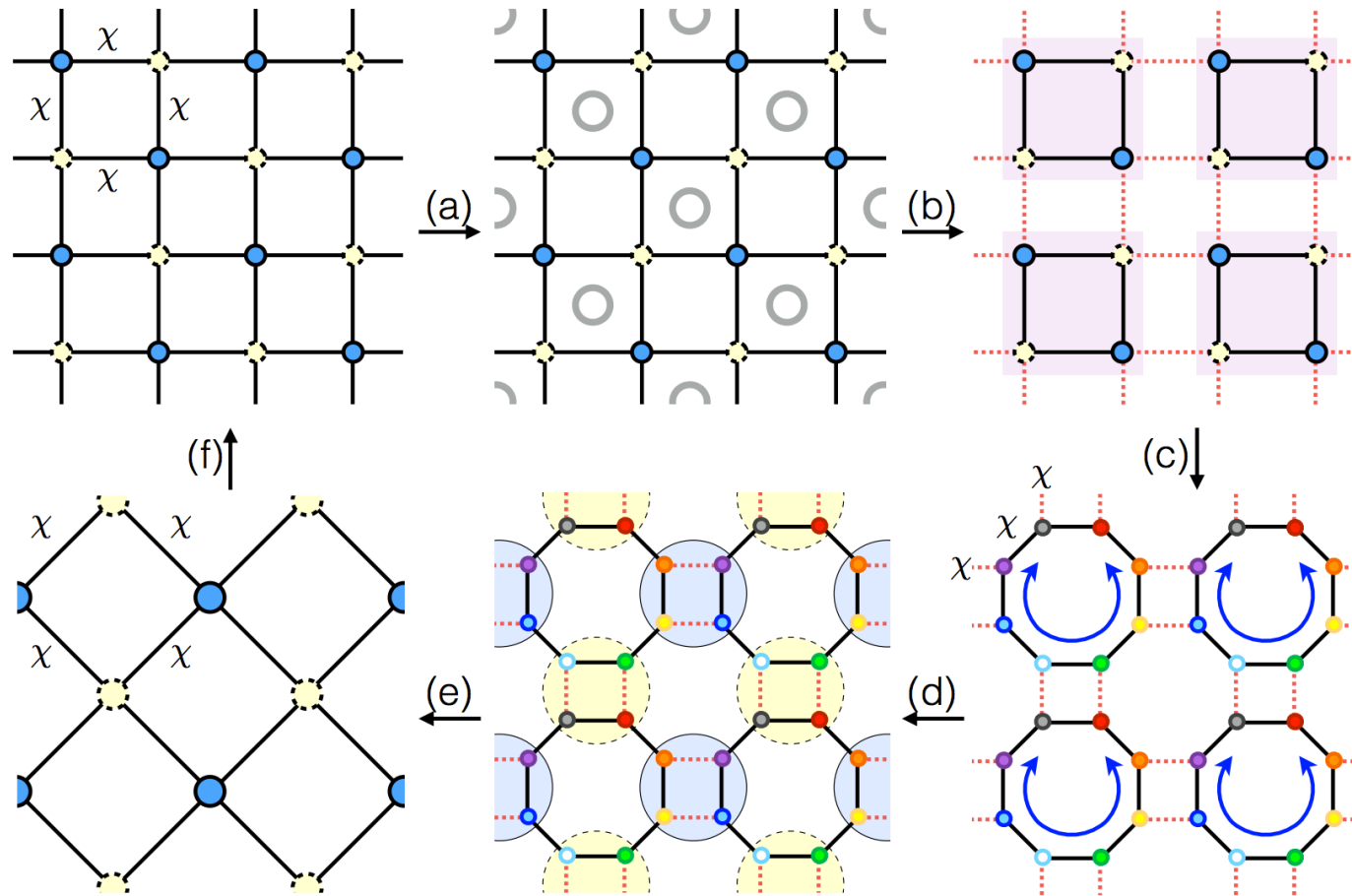
remove short-range entanglement by a local SVD

- Beyond tree level — one loop

LN-TNR + loop optimization
further remove short-range entanglement inside a loop

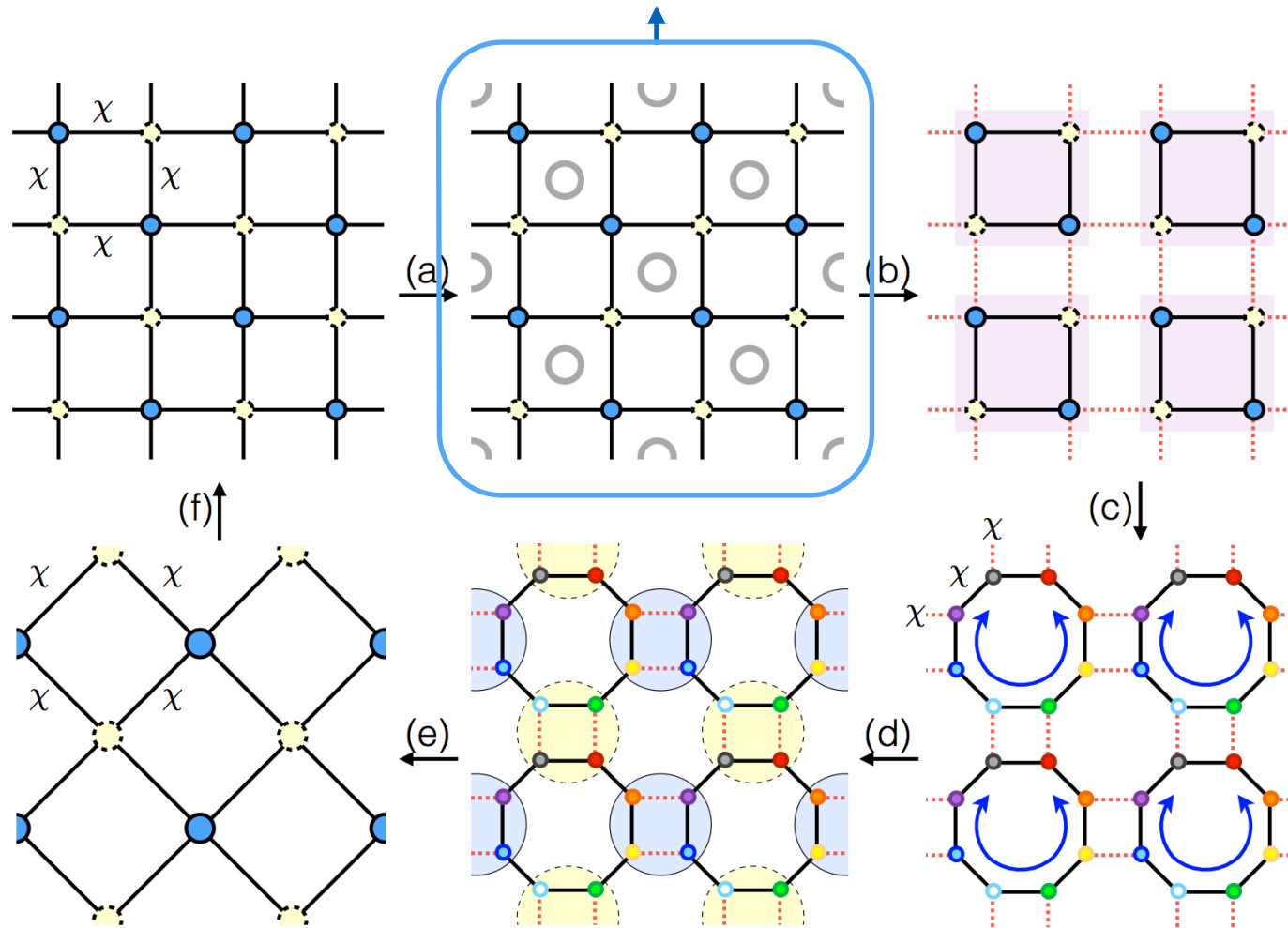
This work !

Algorithms of Loop-TNR



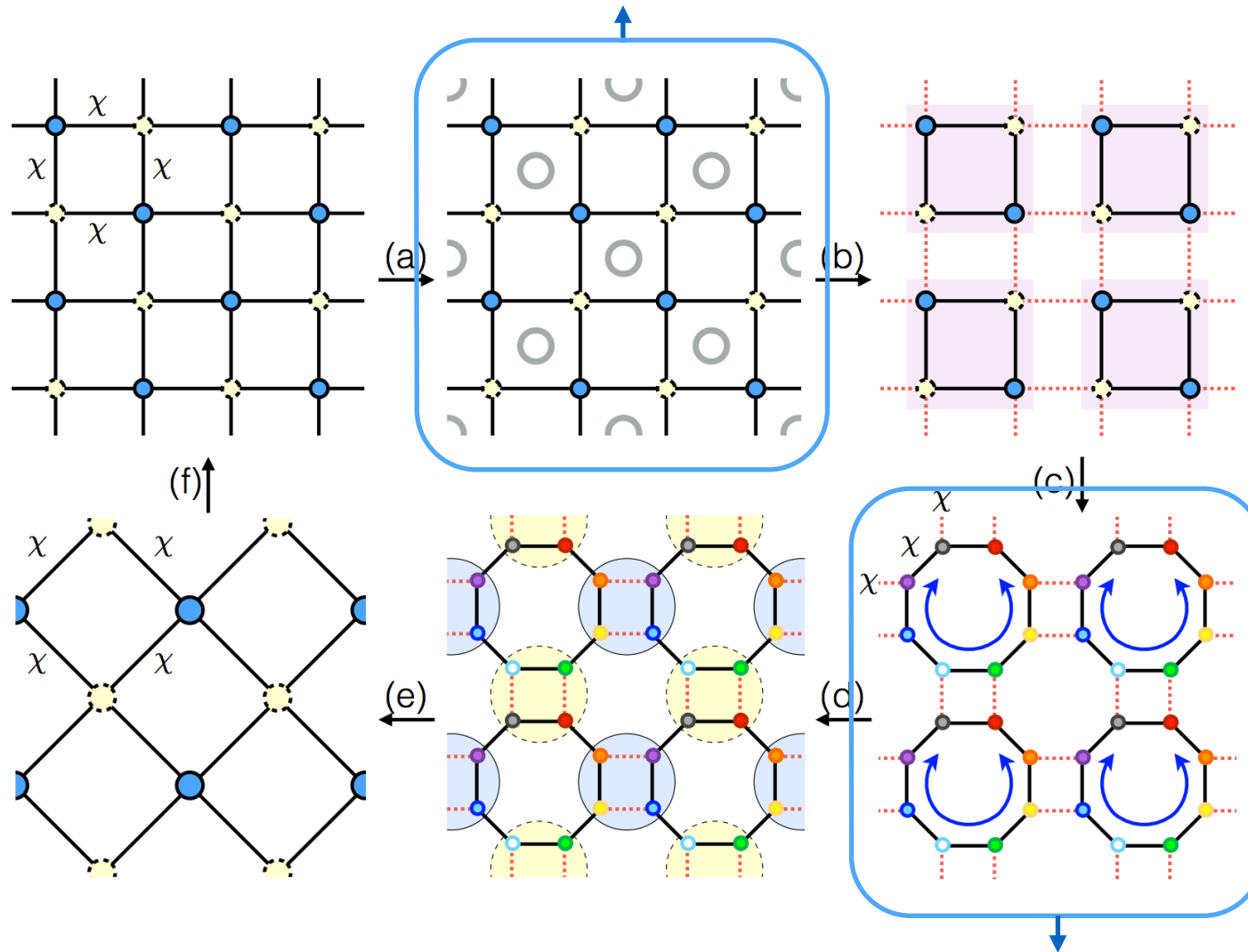
Algorithms of Loop-TNR

Part One: Entanglement filtering



Algorithms of Loop-TNR

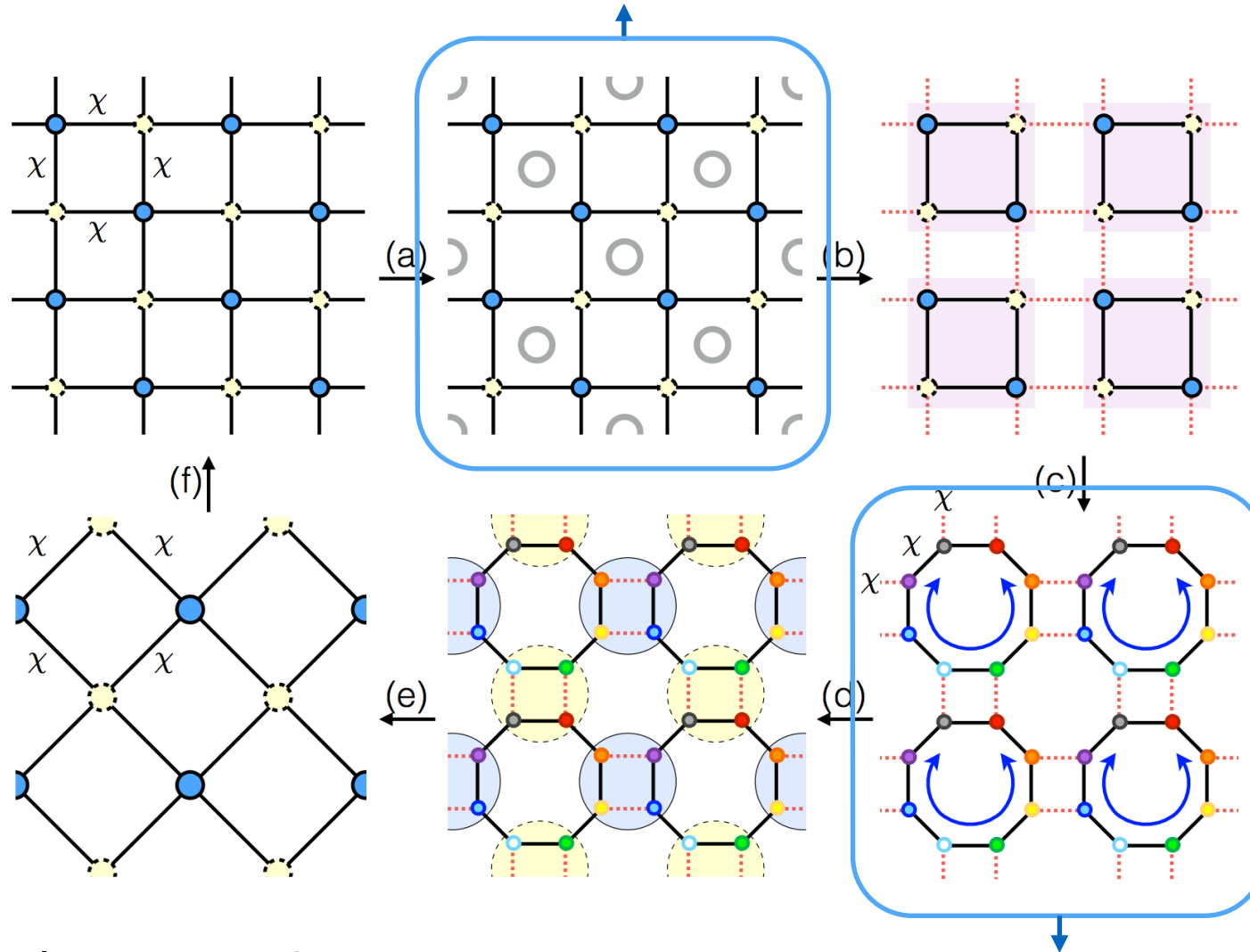
Part One: Entanglement filtering



Part Two: Optimizing tensors on a loop

Algorithms of Loop-TNR

Part One: Entanglement filtering



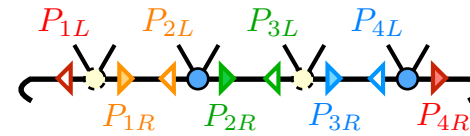
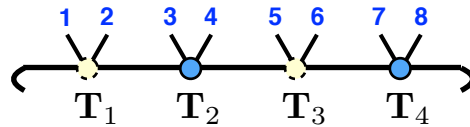
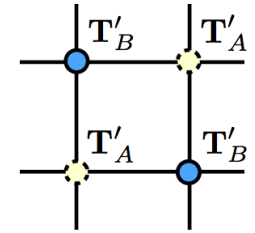
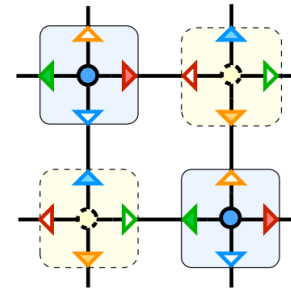
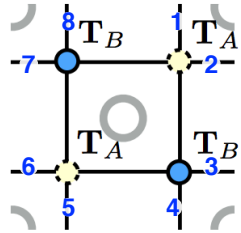
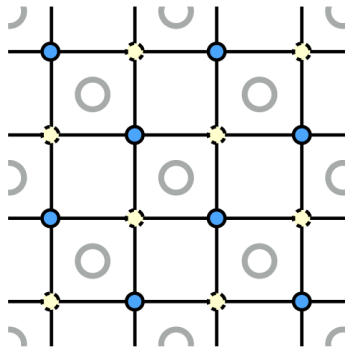
Together: Complete remove short-range entanglement

Part Two: Optimizing tensors on a loop

Part One — Entanglement filtering

How

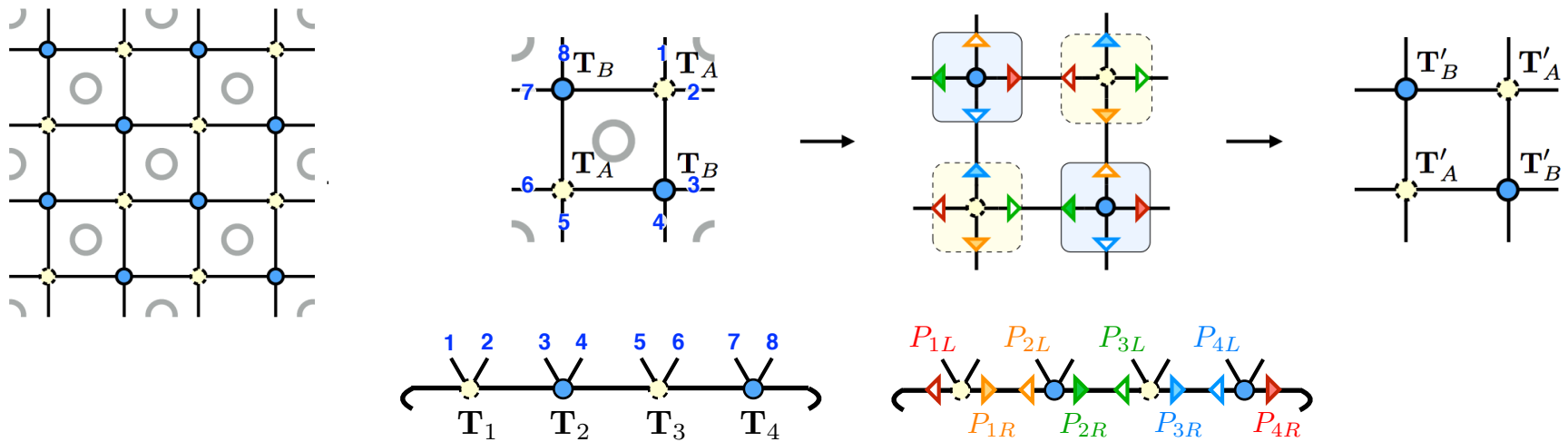
1. Find & insert projectors
2. Define new tensors



Part One — Entanglement filtering

How

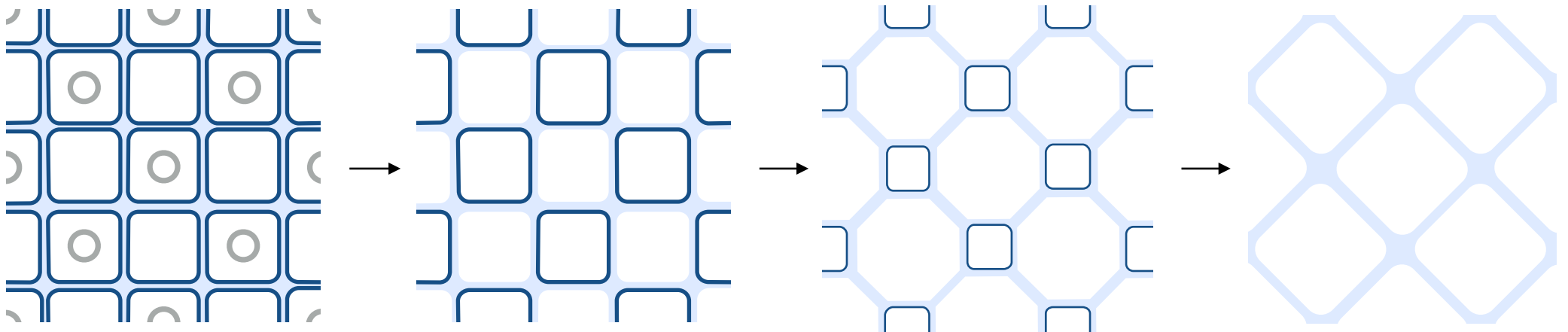
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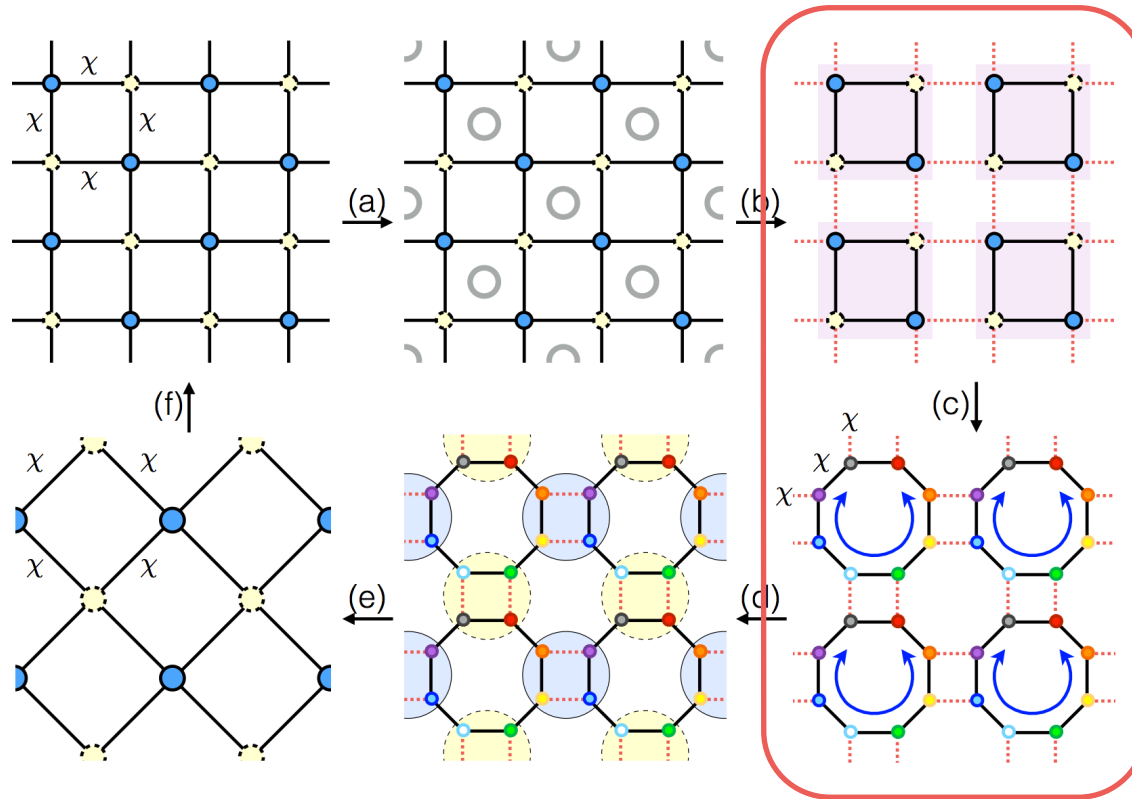
Aim

Remove corner double line (CDL) tensors
Generate local canonical gauge

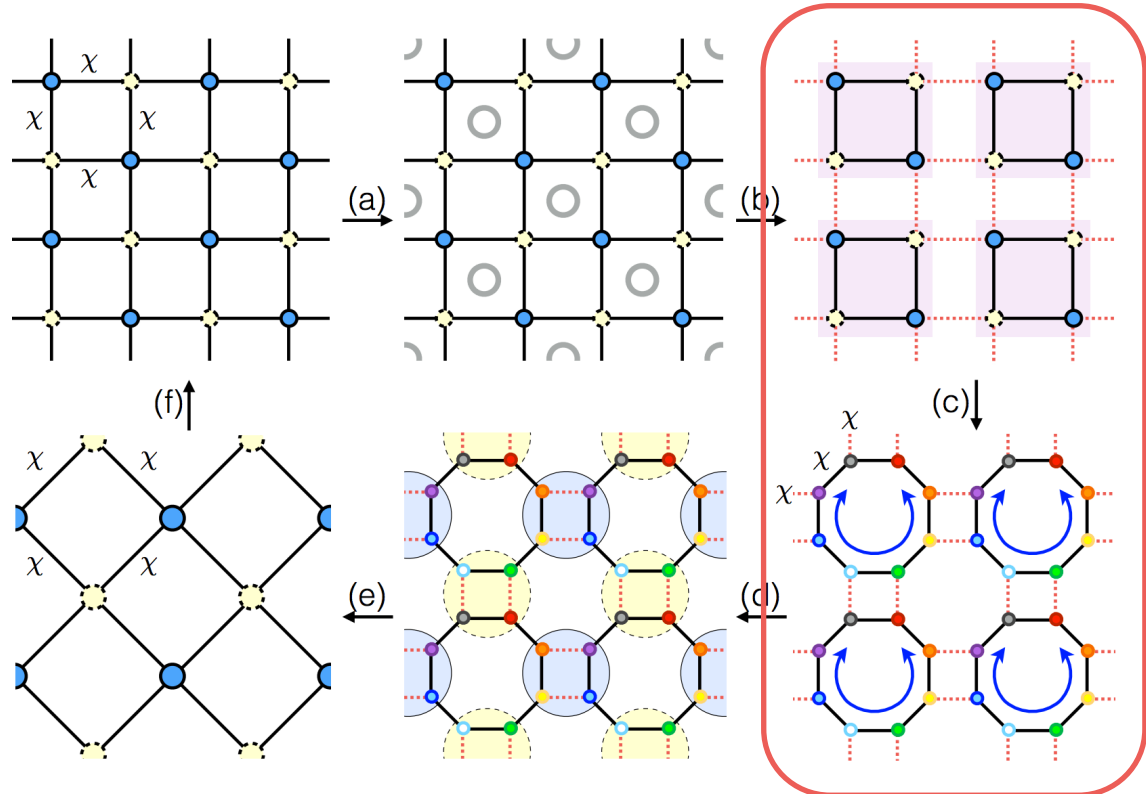
Zheng-Cheng Gu and Xiao-Gang Wen,
Phys. Rev. B **80**, 155131 (2009).



Part Two — Optimizing tensors on a loop



Part Two — Optimizing tensors on a loop



■ LN-TNR

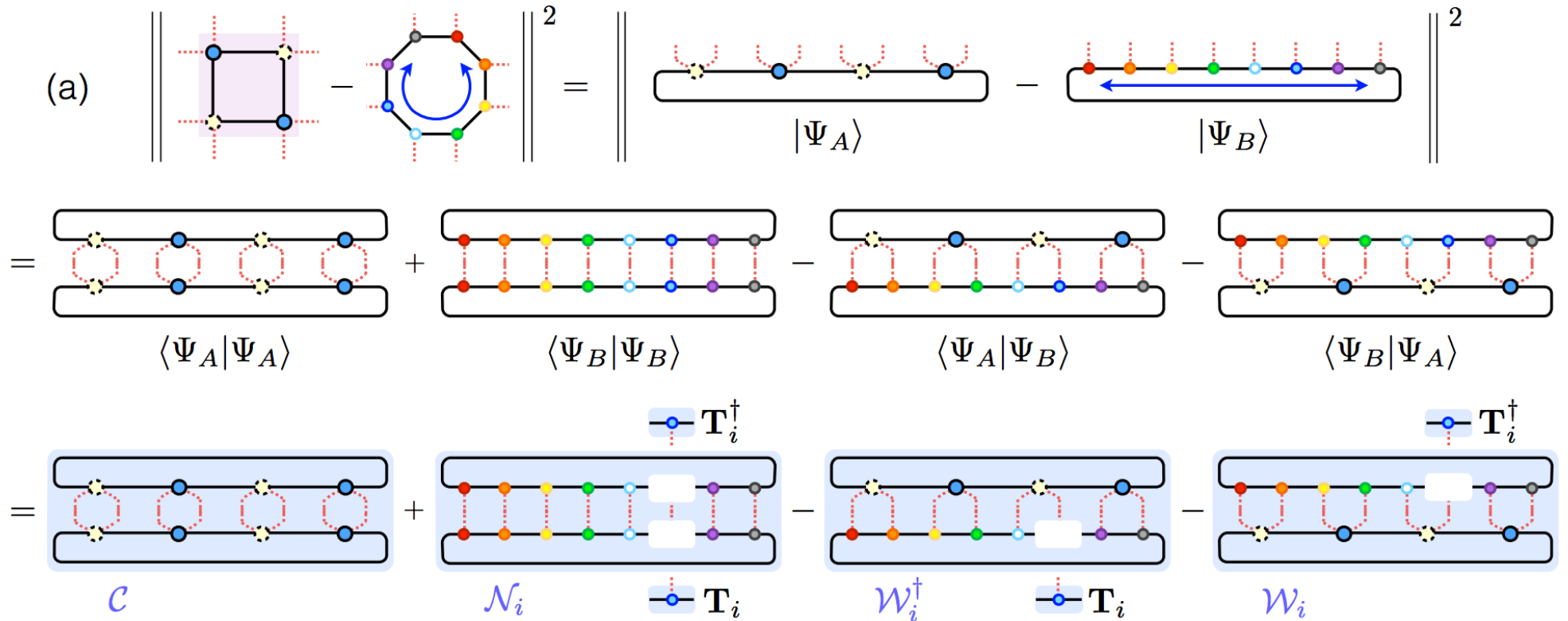
$$\left\| \begin{array}{c} x \\ | \\ x \text{---} \text{---} x \\ | \\ x \end{array} - \begin{array}{c} | \\ 1 \\ | \\ 2 \end{array} \right\|^2, \quad \left\| \begin{array}{c} x \\ | \\ x \end{array} - \begin{array}{c} | \\ 3 \\ | \\ 4 \end{array} \right\|^2$$

cost function

■ Loop-TNR

$$\left\| \begin{array}{c} T_4 \quad T_1 \\ | \quad | \\ T_3 \quad T_2 \end{array} - \begin{array}{c} 8 \quad 1 \\ | \quad | \\ 7 \quad 2 \\ | \quad | \\ 6 \quad 3 \\ | \quad | \\ 5 \quad 4 \end{array} \right\|^2 = \left\| \begin{array}{c} \text{---} \text{---} \text{---} \text{---} \\ | \quad | \quad | \quad | \\ \text{---} \text{---} \text{---} \text{---} \end{array} - \begin{array}{c} \text{---} \text{---} \text{---} \text{---} \\ | \quad | \quad | \quad | \\ \text{---} \text{---} \text{---} \text{---} \end{array} \right\|^2$$

Part Two — Optimizing tensors on a loop



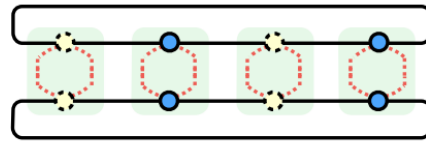
$$\begin{aligned}
 f(\mathbf{T}_i) &= \left\| |\Psi_A\rangle - |\Psi_B\rangle \right\|^2 = \langle \Psi_A | \Psi_A \rangle + \langle \Psi_B | \Psi_B \rangle - \langle \Psi_A | \Psi_B \rangle - \langle \Psi_B | \Psi_A \rangle \\
 &= \mathcal{C} + \mathbf{T}_i^\dagger \mathcal{N}_i \mathbf{T}_i - \mathcal{W}_i^\dagger \mathbf{T}_i - \mathbf{T}_i^\dagger \mathcal{W}_i,
 \end{aligned}$$

solve the linear equation

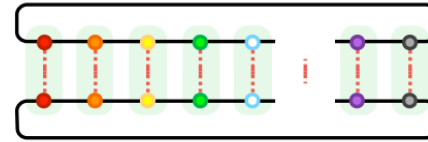
$$\mathcal{N}_i \mathbf{T}_i = \mathcal{W}_i.$$

Part Two — Optimizing tensors on a loop

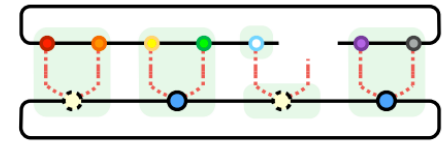
(b) contraction order



C



N_i



W_i

computational cost

$$\mathcal{O}(7\chi^6)$$

$$\mathcal{O}(6\chi^6 + 7\chi^5)$$

$$\mathcal{O}(6\chi^6 + 4\chi^5)$$

memory cost

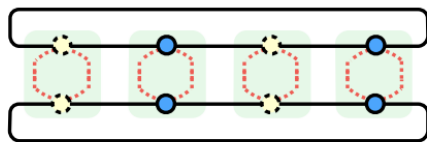
$$\mathcal{O}(\chi^4)$$

$$\mathcal{O}(\chi^4)$$

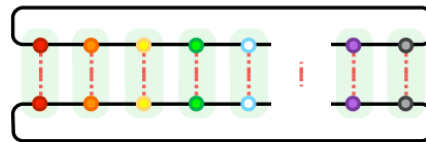
$$\mathcal{O}(\chi^4)$$

Part Two — Optimizing tensors on a loop

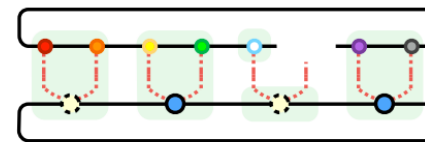
(b) contraction order



C



N_i



W_i

computational cost

$$\mathcal{O}(7\chi^6)$$

$$\mathcal{O}(6\chi^6 + 7\chi^5)$$

$$\mathcal{O}(6\chi^6 + 4\chi^5)$$

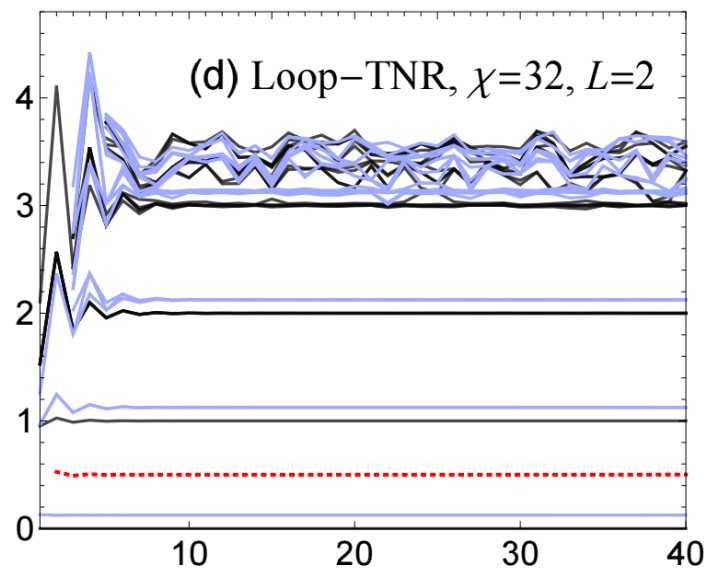
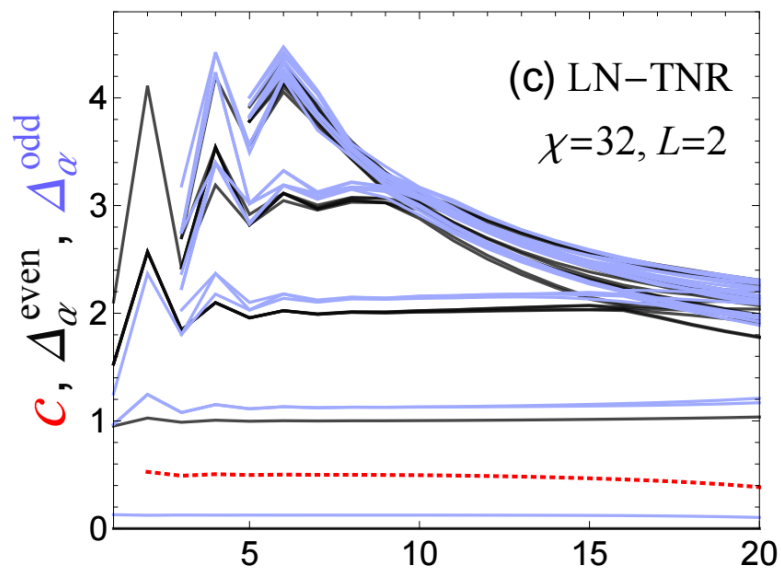
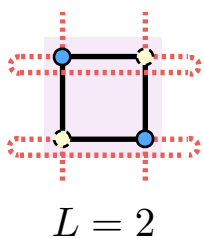
memory cost

$$\mathcal{O}(\chi^4)$$

$$\mathcal{O}(\chi^4)$$

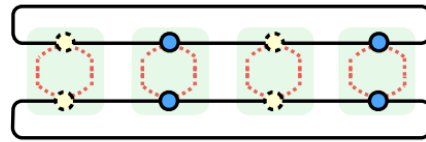
$$\mathcal{O}(\chi^4)$$

Loop-TNR Results

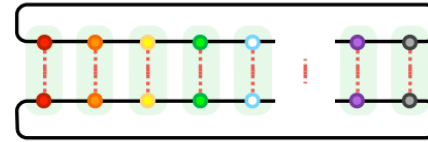


Part Two — Optimizing tensors on a loop

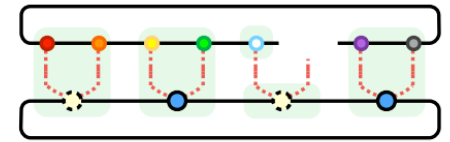
(b) contraction order



C



N_i



W_i

computational cost

$$\mathcal{O}(7\chi^6)$$

$$\mathcal{O}(6\chi^6 + 7\chi^5)$$

$$\mathcal{O}(6\chi^6 + 4\chi^5)$$

memory cost

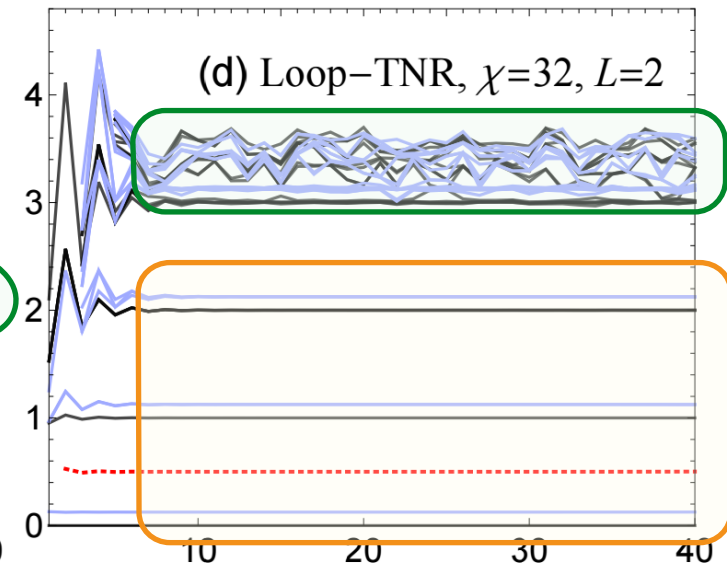
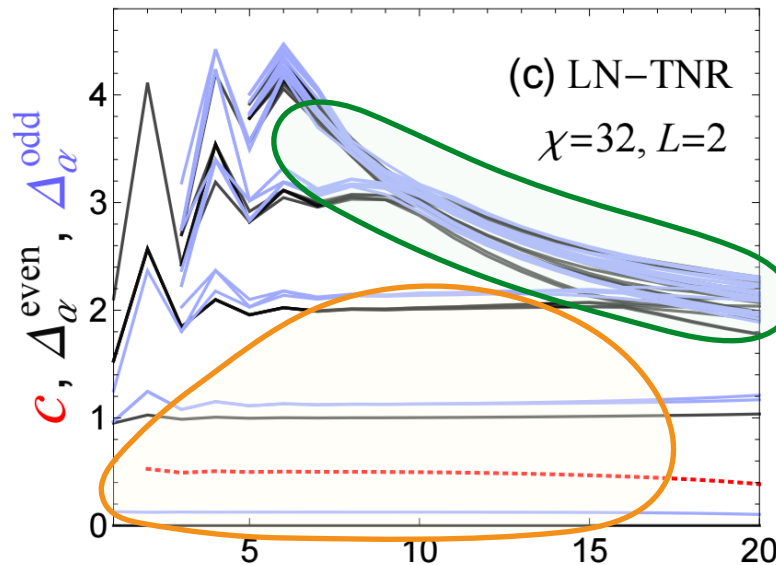
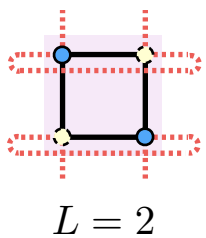
$$\mathcal{O}(\chi^4)$$

$$\mathcal{O}(\chi^4)$$

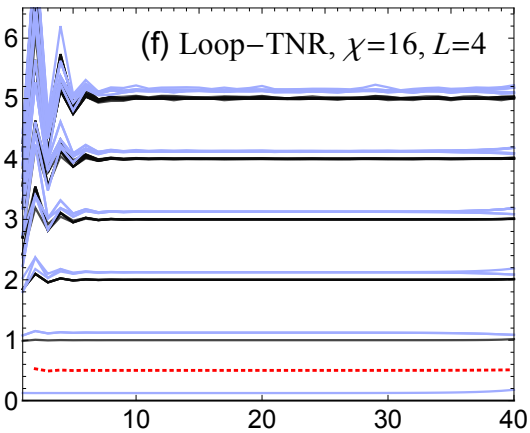
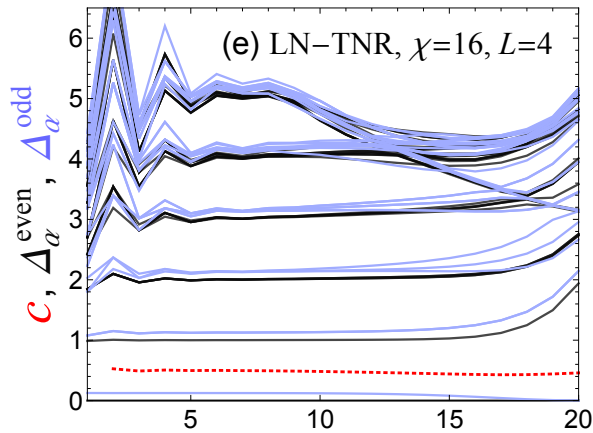
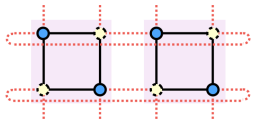
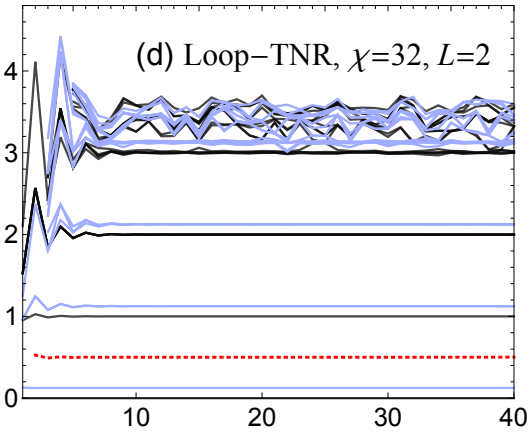
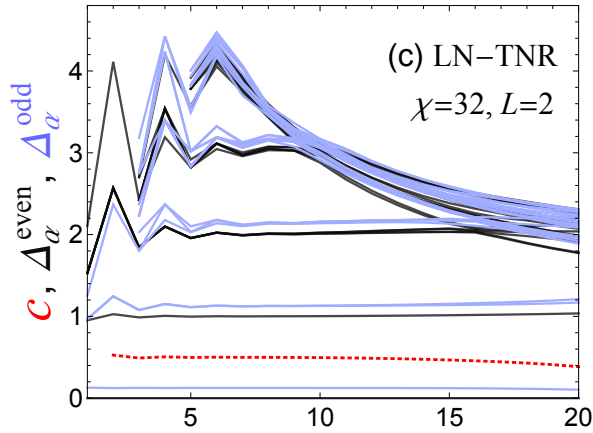
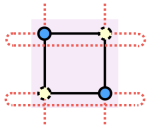
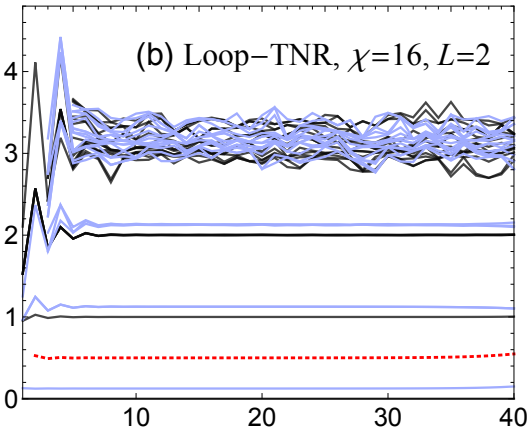
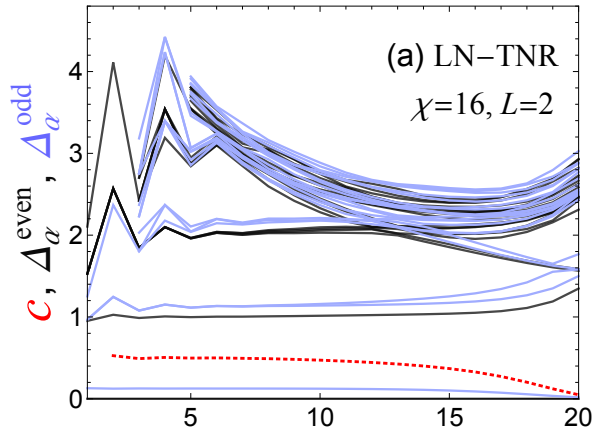
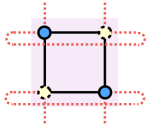
$$\mathcal{O}(\chi^4)$$

Loop-TNR Results

- scaling dimensions does not change with scale \sim scale invariance
- a clear gap between high-level parts and low-level parts



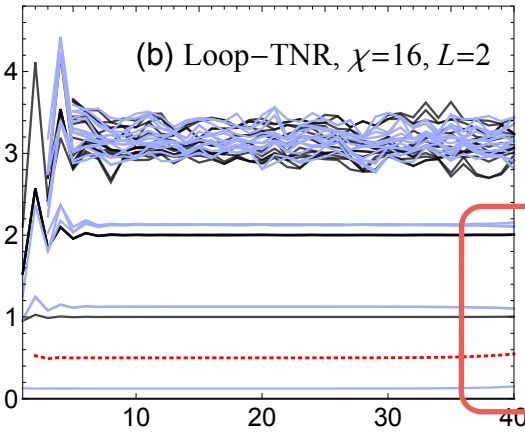
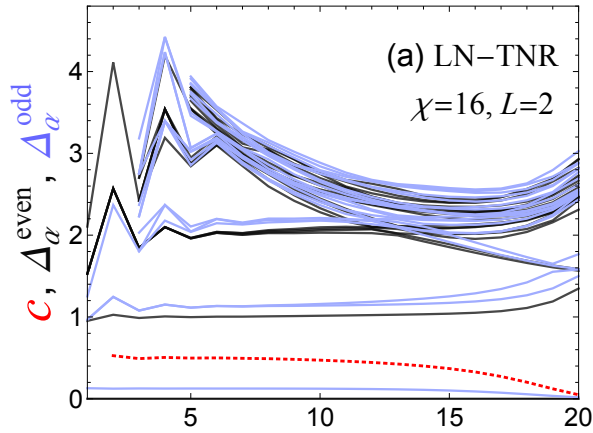
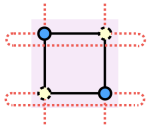
Stability



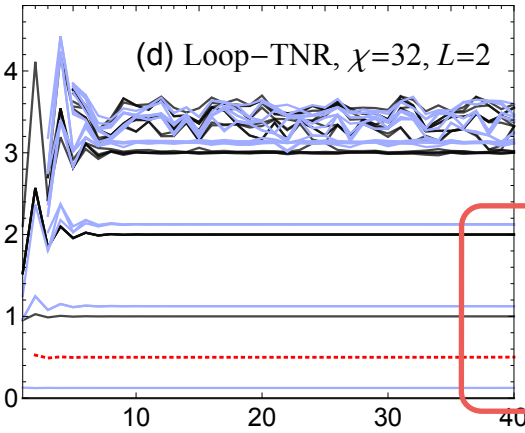
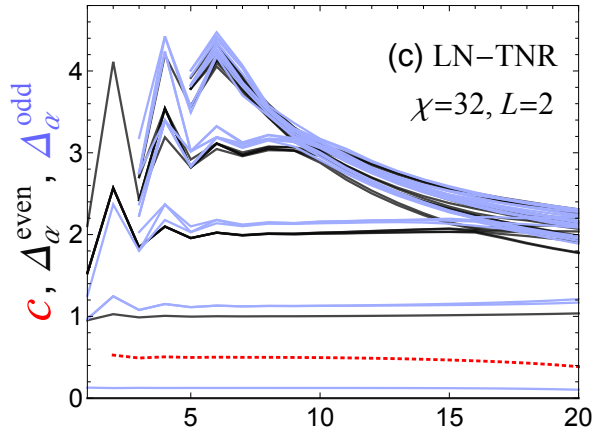
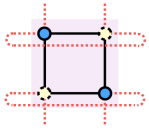
Iteration step

Iteration step

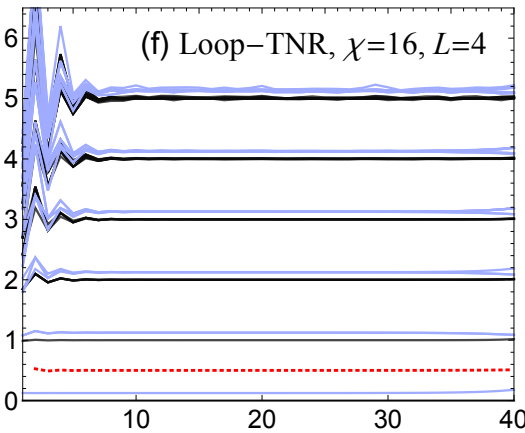
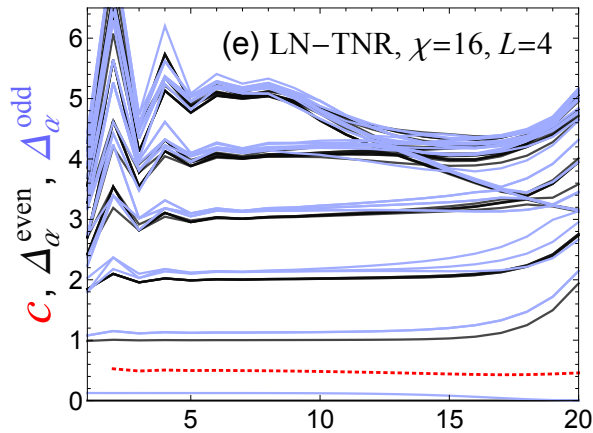
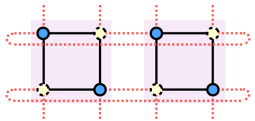
Stability



$\chi = 16$
remain accurate up to 40
iteration steps



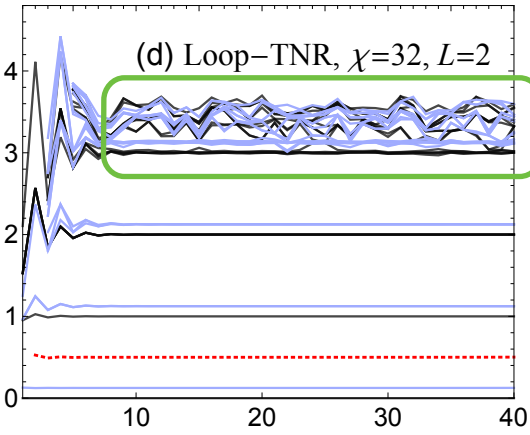
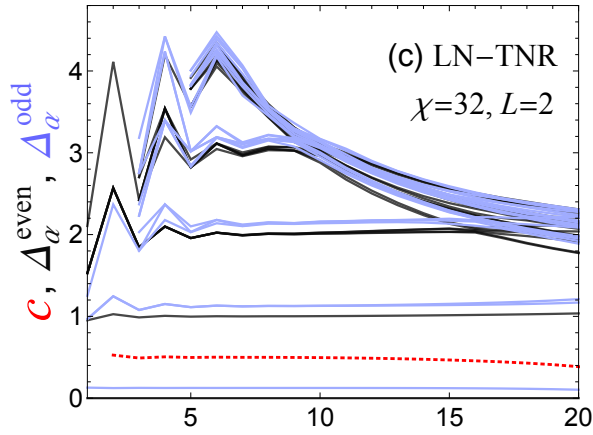
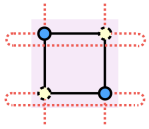
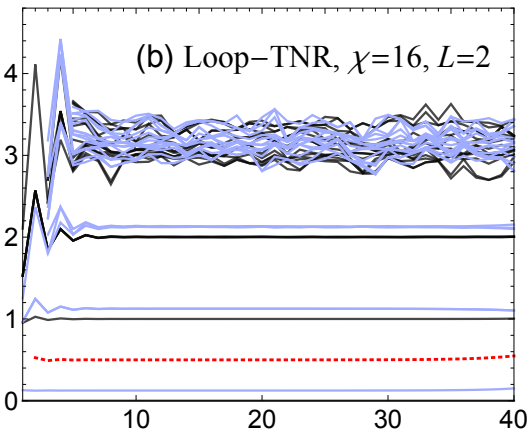
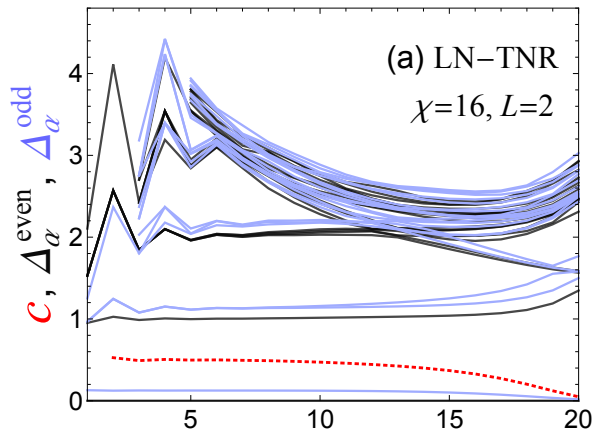
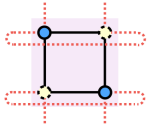
even longer for $\chi = 32$
the proper RG flow last
longer for larger χ



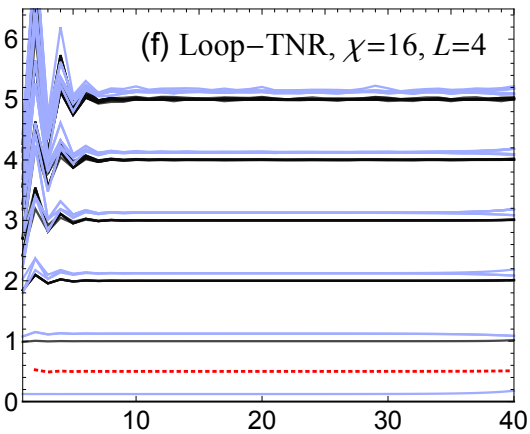
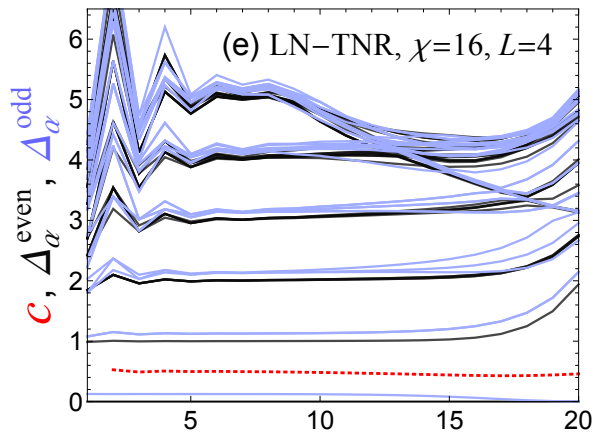
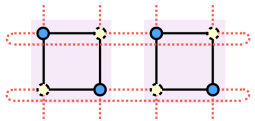
Iteration step

Iteration step

Stability



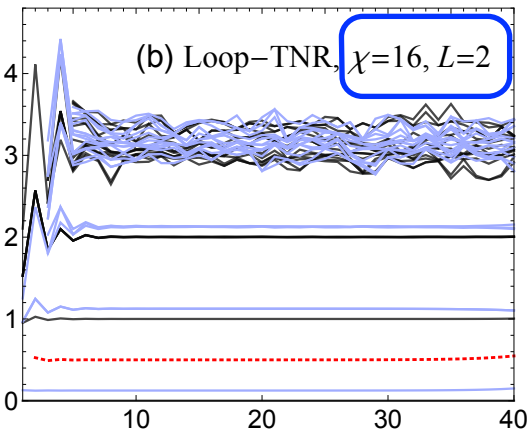
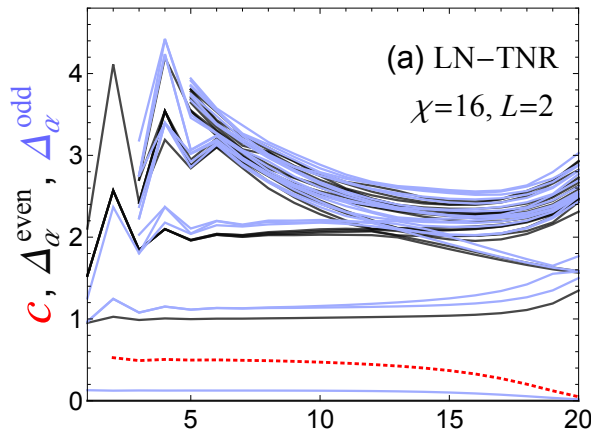
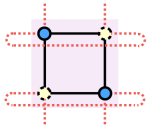
increasing χ , more scaling dimensions can be resolved



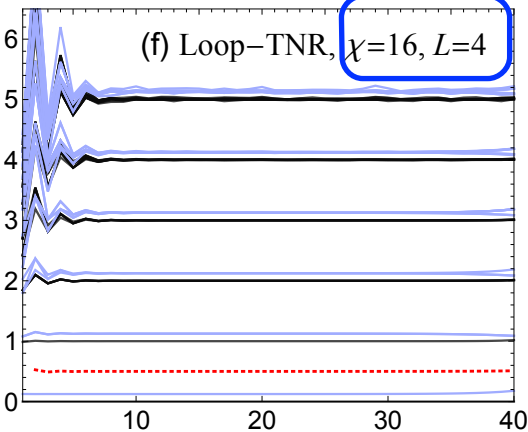
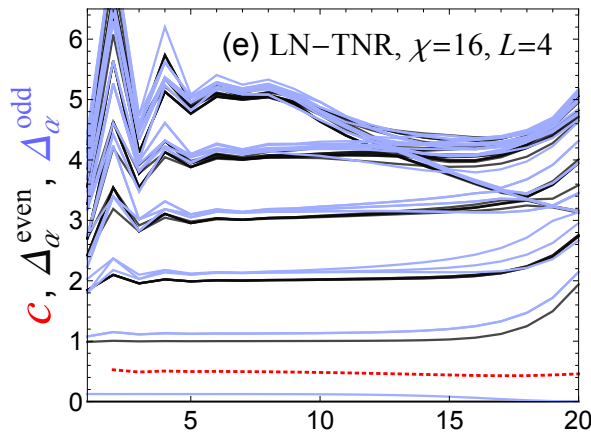
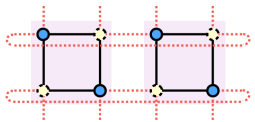
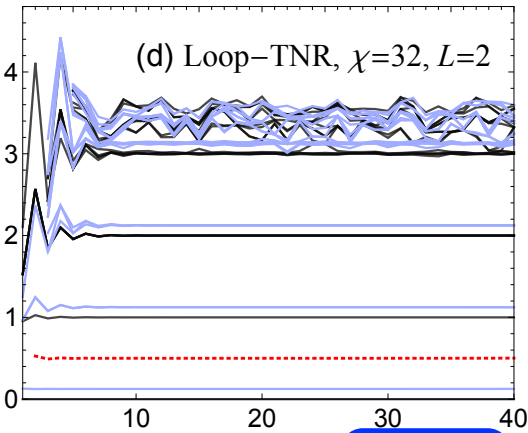
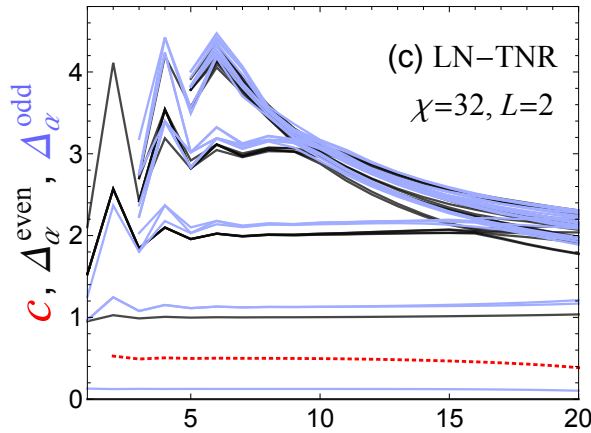
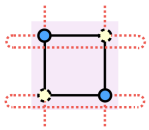
Iteration step

Iteration step

Stability



effectively, $\chi = 16$

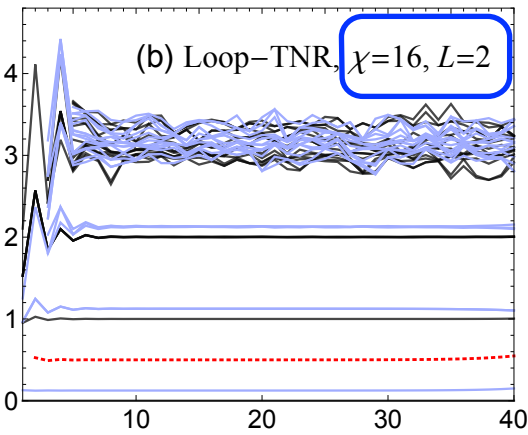
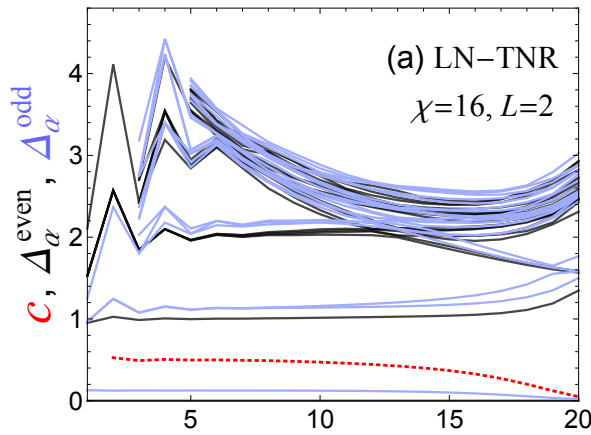
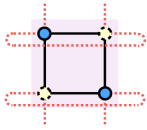


effectively, $\chi = 16^2 = 256$
much more scaling
dimensions can be read off
accuracy is higher

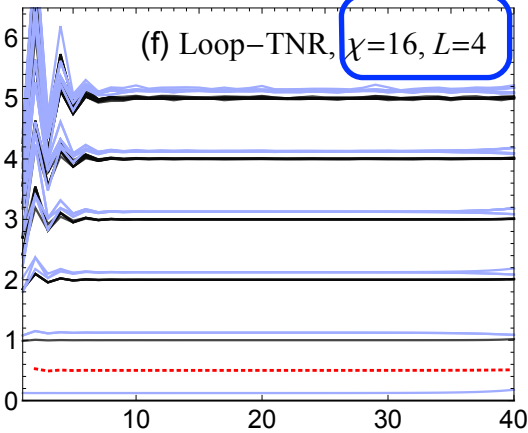
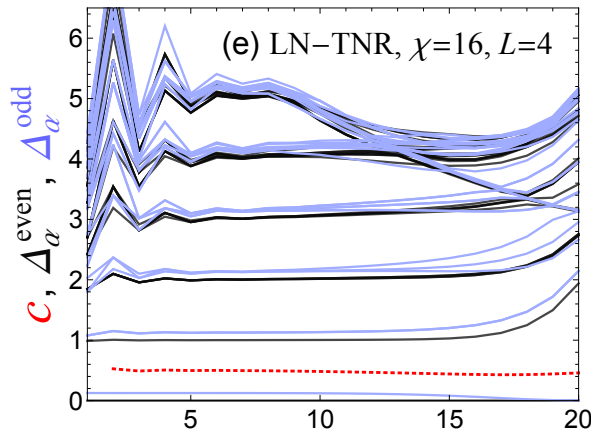
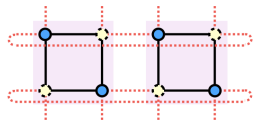
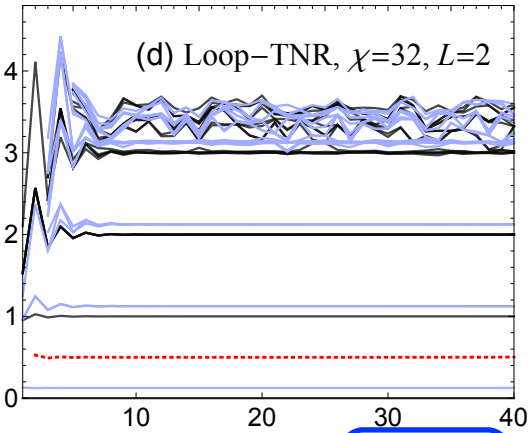
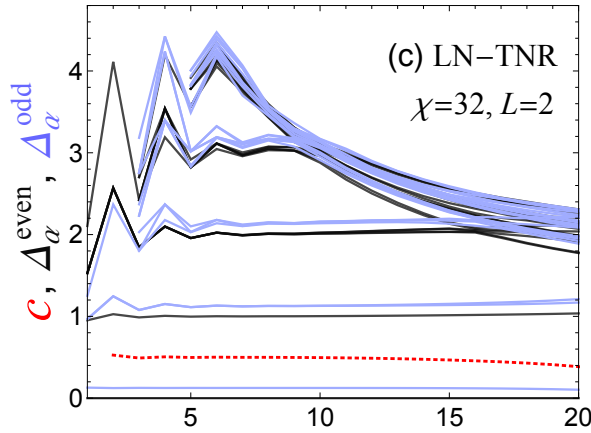
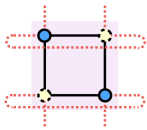
Iteration step

Iteration step

Stability



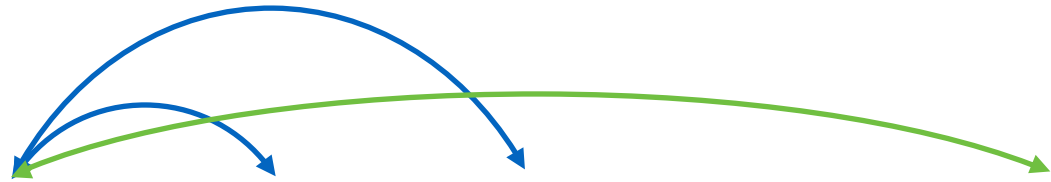
effectively, $\chi = 16$



effectively, $\chi = 16^2 = 256$
 much more scaling
 dimensions can be read off
 accuracy is higher

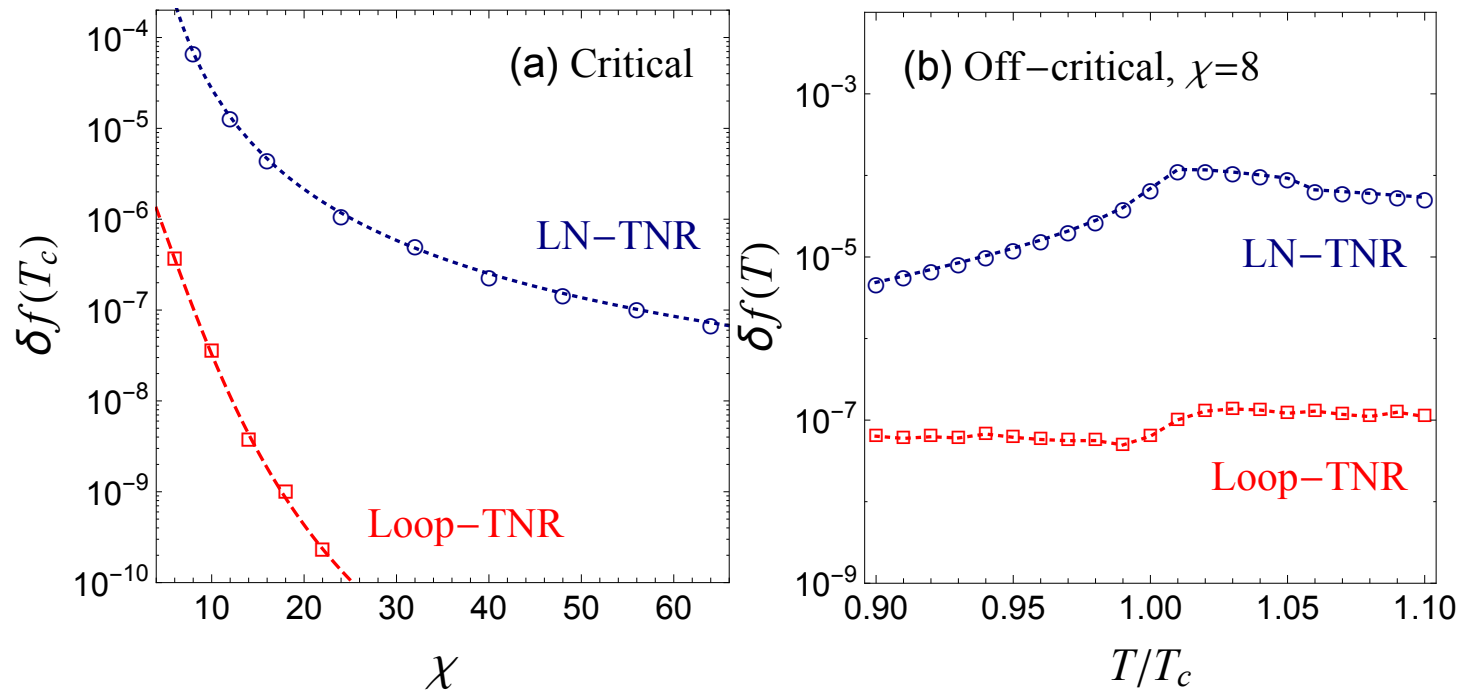
infinite $\chi \sim$ infinite
 dimensional fixed point tensor
 described by Ising CFT

Accuracy



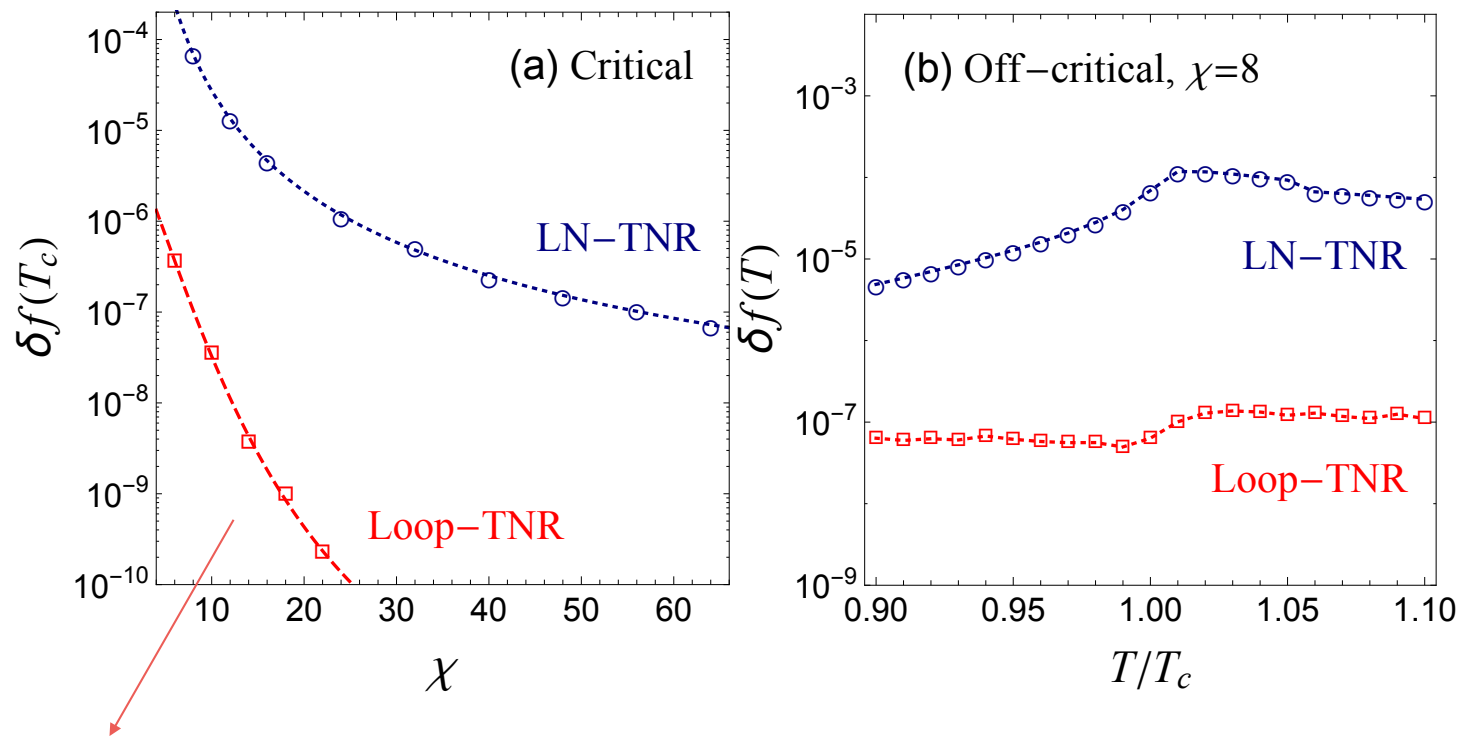
Exact	LN-TNR $\chi = 64$ $L = 1$ 2^{11} spins	LN-TNR $\chi = 64$ $L = 2$ 2^{11} spins	Loop-TNR $\chi = 16$ $L = 2$ 2^{18} spins	Loop-TNR $\chi = 24$ $L = 2$ 2^{18} spins	Loop-TNR $\chi = 16$ $L = 4$ 2^{18} spins	Loop-TNR $\chi = 24$ $L = 4$ 2^{18} spins	EV-TNR [50] $\chi = 24$ $L = 2$ 2^{18} spins
$c = 0.5$	0.49946958	0.49970058	0.50001491	0.50000165	0.50009255	0.50008794	0.50001
$\sigma = 0.125$	0.12504027	0.12500837	0.12500528	0.12500011	0.12501117	0.12499789	0.1250004
$\epsilon = 1$	1.00028269	0.99996784	1.00000566	1.00000601	0.99999403	1.00000507	1.00009
1.125	1.12368834	1.12444247	1.12495187	1.12499400	1.12498755	1.12500559	1.12492
1.125	1.12394625	1.12450246	1.12510600	1.12500464	1.12498755	1.12500559	1.12510
2	1.92334948	1.99811859	2.00000743	1.99970911	1.99999517	2.00000985	1.99922
2	1.96264143	1.99815644	2.00066117	2.00016629	1.99999517	2.00000985	1.99986
2	1.97496787	1.99868822	2.00066117	2.00031103	2.00002744	2.00001690	2.00006
2	2.00274974	1.99948966	2.00586886	2.00131384	2.00006203	2.00002745	2.00168

Relative error of the free energy per site



Relative error of the free energy per site

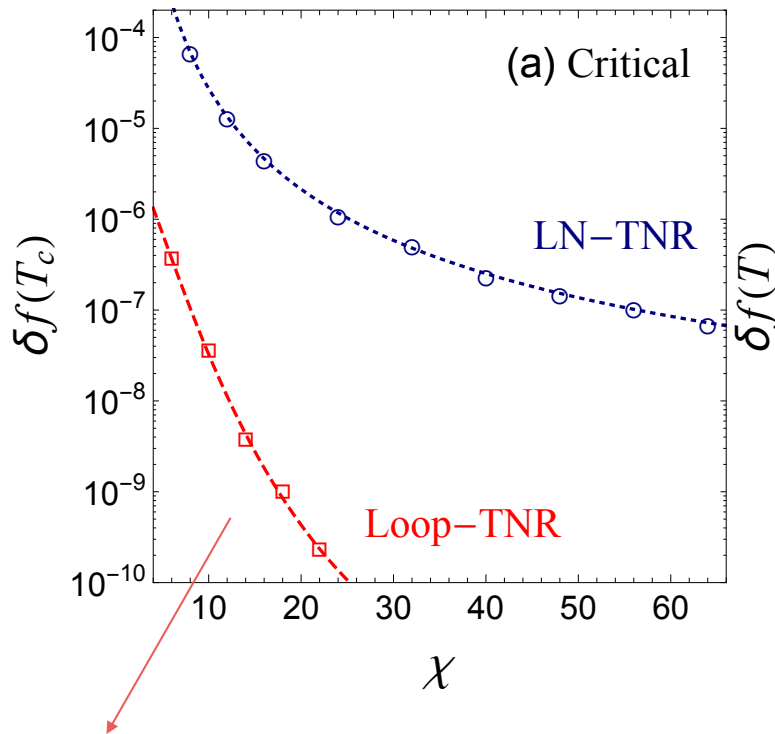
- At critical point, the error of Loop-TNR decays much faster than the error of LN-TNR



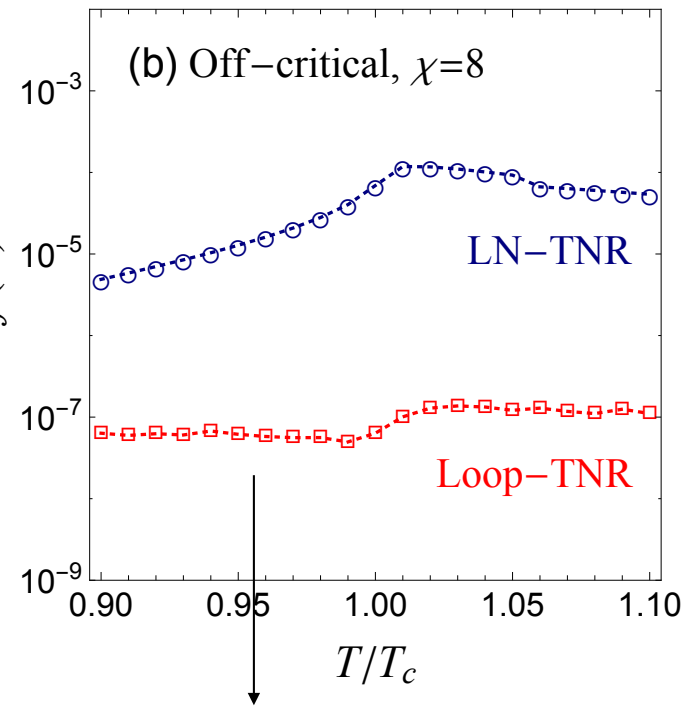
almost decays exponentially with bond dimension

Relative error of the free energy per site

- At critical point, the error of Loop-TNR decays much faster than the error of LN-TNR

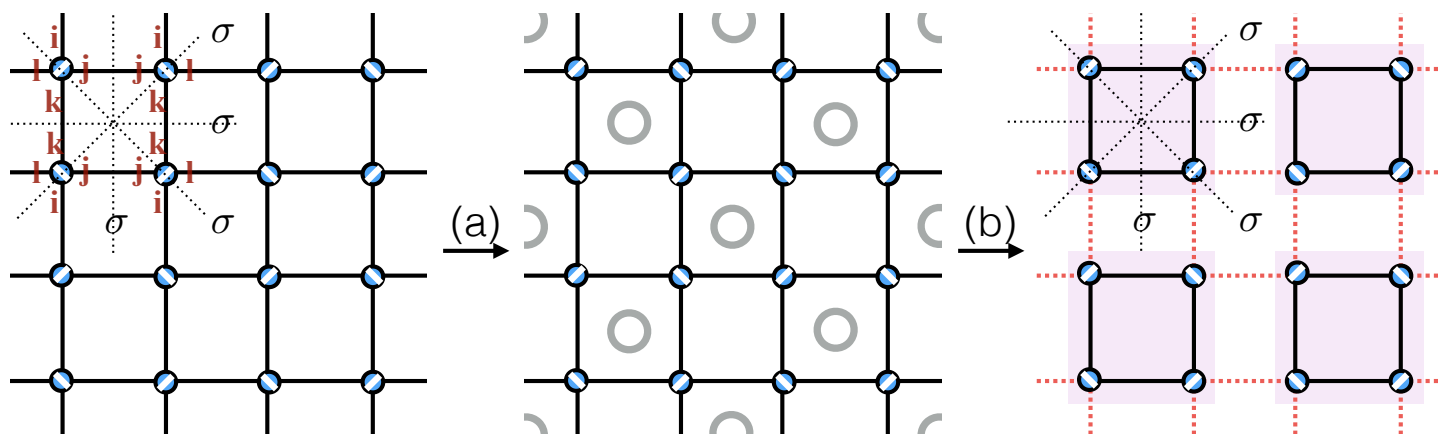


almost decays exponentially with bond dimension

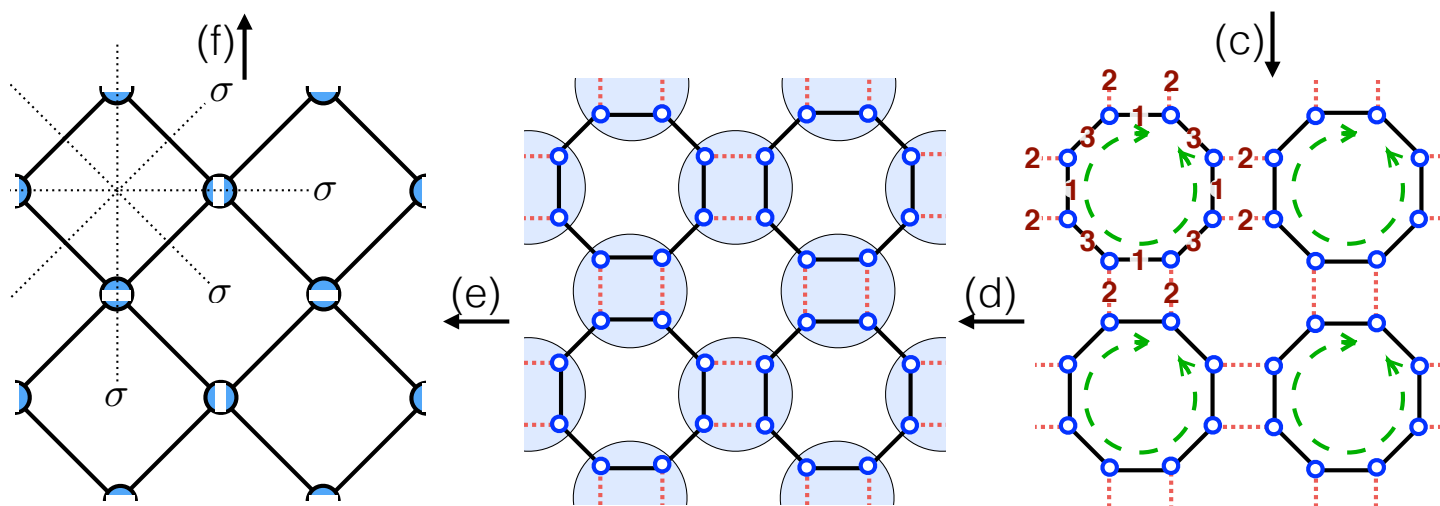


- At off-critical points, the errors almost remains a constant for all temperatures near the critical point

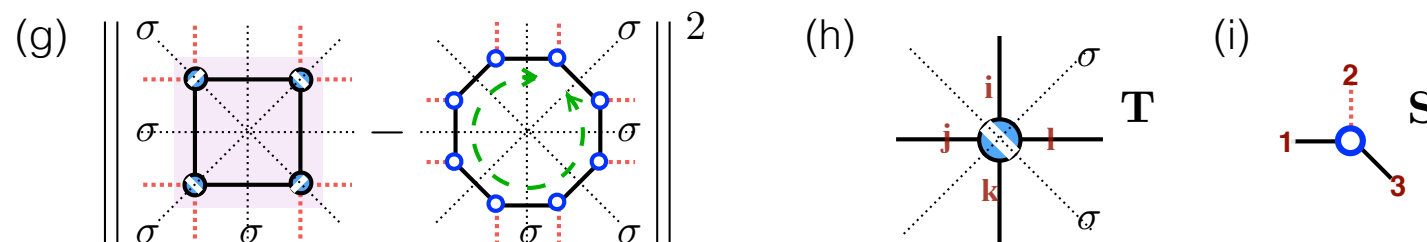
Loop TRG with reflection symmetry



4 axis of symmetry
global C4 symmetry

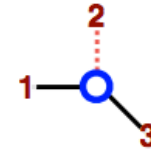


fix the gauge

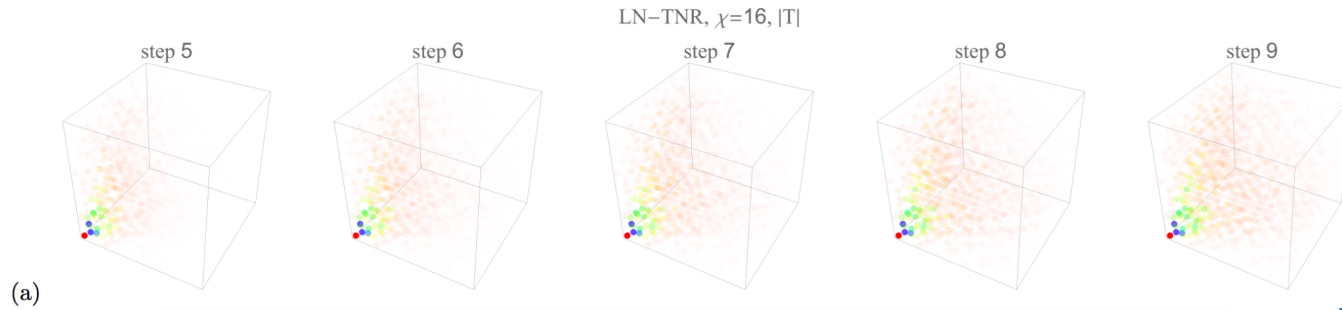


ideal case:
explicitly
invariant fixed
point tensor S

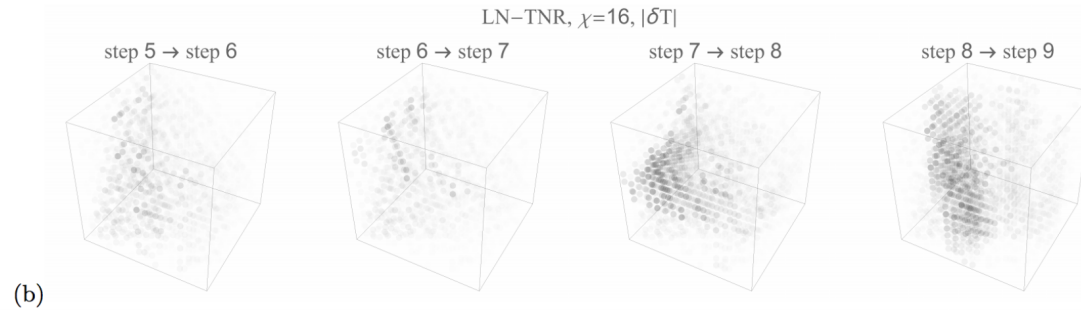
Fixed point tensors



LN-TNR
absolute value

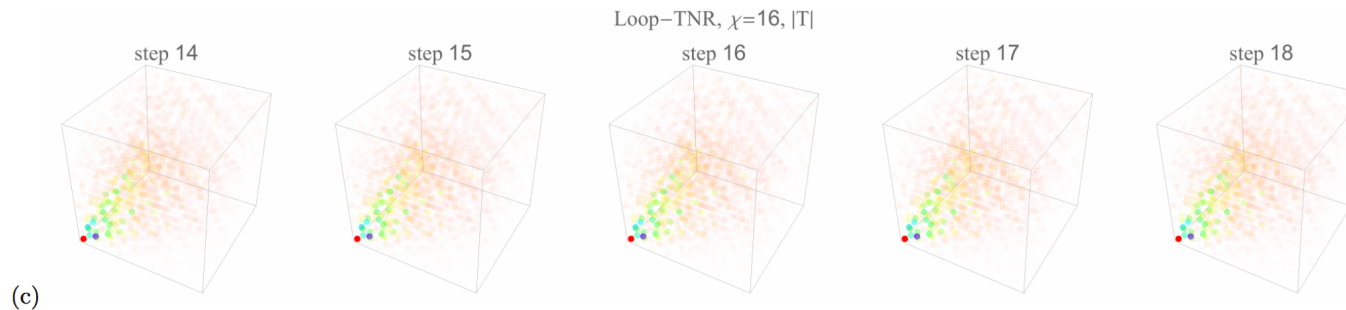


LN-TNR
absolute difference

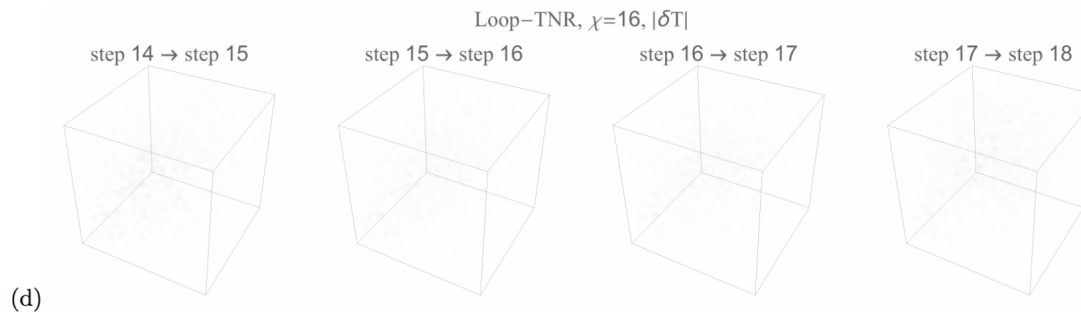


low-index parts
are approximately
invariant

Loop-TNR
absolute value



Loop-TNR
absolute difference



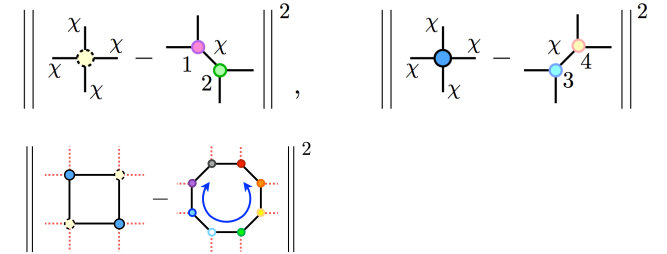
all tensor elements
are nearly invariant
 \sim scale invariance

Summary

Summary

Real space TNR

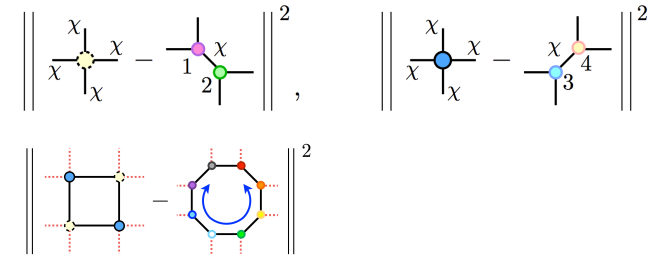
- Tree level: LN-TNR (Levin & Nave, 2007)
- One loop: GW-TNR, EV-TNR, Loop-TNR


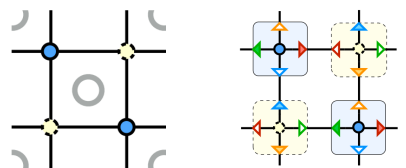
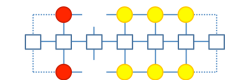
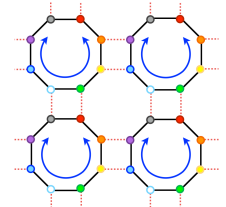
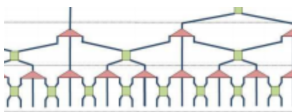
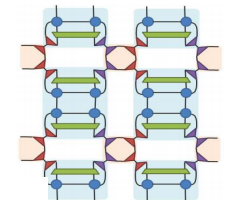


Summary

Real space TNR

- Tree level: LN-TNR (Levin & Nave, 2007)
- One loop: GW-TNR, EV-TNR, Loop-TNR



1D algorithm	LN-TNR + 1D algorithm
 <p>iTEBD</p>	<p>GW-TNR (Gu & Wen, 2009) Loop-TNR Part One</p> 
 <p>DMRG Variational MPS</p>	<p>Loop-TNR Part Two</p> 
 <p>MERA</p>	<p>EV-TNR (Evenbly & Vidal, 2014)</p> 

Thank you!