

Tensor network study of CP(1) model with theta-term

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(Kanazawa U.)

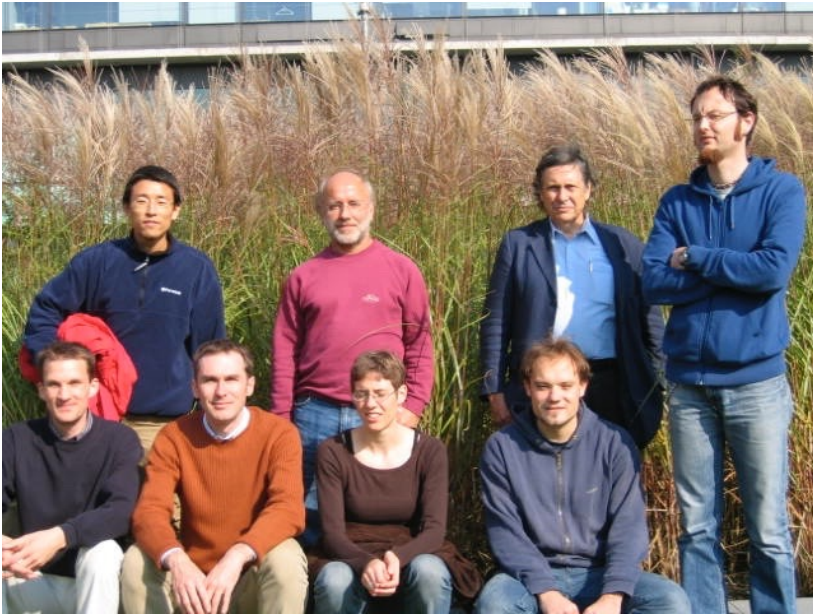
In collaboration with H. Aizawa and Y. Yoshimura



German Japanese Workshop 2024

25-27.09.2024 @Mainz

2005-2008 Postdoc time



Contents

- Introduction of tensor networks
- Application to CP(1) + θ model

Tensor networks (TN)

- Why TN? Answer: **No sign-problem**

Real-time, Finite density, θ -term, ...

- What's TN?

we address later

$$\text{Diagram} = \sum_{\dots, i, j, k, l, \dots} \dots T_{ijkl} T_{mnio} \dots$$

- TN is used as a representation of wave function or **partition function (path integral)**

How to compute Z using TN

$$Z \equiv \int [d\phi] e^{-S[\phi]}$$

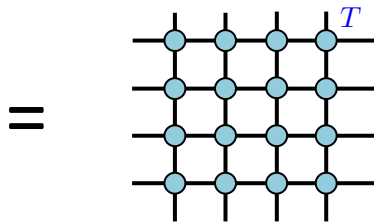
local interaction

How to compute Z using TN

local interaction

$$Z \equiv \int [d\phi] e^{-S[\phi]} = \sum_{\dots, i, j, k, l, \dots} \dots T_{ijkl} T_{mnio} \dots$$

Periodic BC



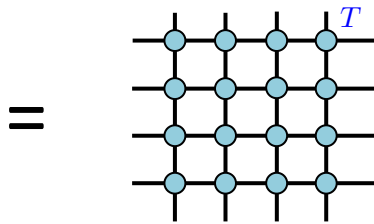
- Non-trivial step, but
- OK for scalar, gauge, and fermion fields as long as the interaction is local

- Scalar : [Shimizu 2012](#), [Sakai+ 2018](#)
- Gauge: [Meurice+ 2013](#), [Nishimura+ 2021](#), [Fukuma+ 2021](#)
- Fermions: [Shimizu+Kuramashi 2014](#)

How to compute Z using TN

$$Z \equiv \int [d\phi] e^{-S[\phi]} = \sum_{\dots, i, j, k, l, \dots} \cdots T_{ijkl} T_{mnio} \cdots$$

Periodic BC



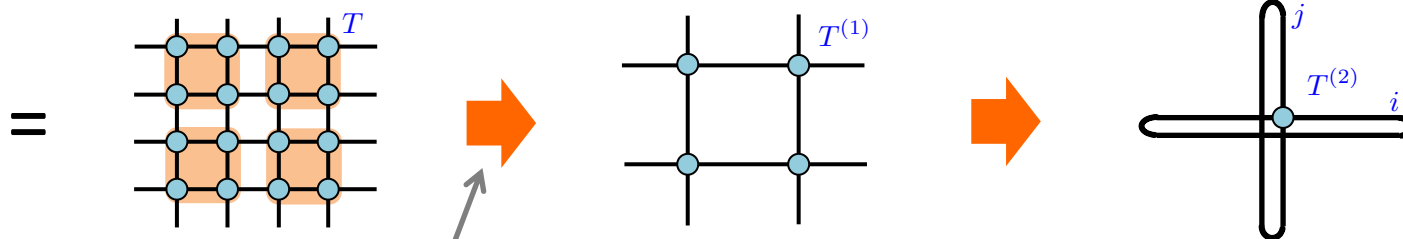
$$\text{cost} \propto \chi^{2V} \quad \text{for} \quad 1 \leq i, j, \dots \leq \chi$$

Bond dimension

How to compute Z using TN

$$Z \equiv \int [d\phi] e^{-S[\phi]} = \sum_{\dots, i, j, k, l, \dots} \cdots T_{ijkl} T_{mnio} \cdots$$

Periodic BC



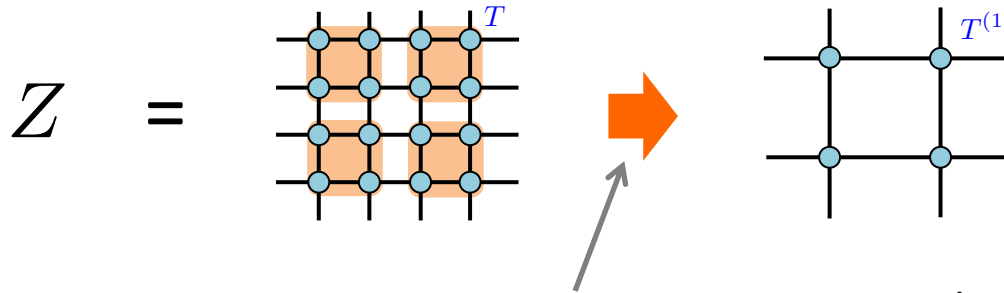
Coarse-graining like spin-blocking

$$Z \approx \sum_{i, j} T_{ijij}^{(n)}$$

Information compression using
singular value decomposition (SVD)

Coarse-graining

Tensor renormalization group (TRG)
Levin+Nave 2007



Information compression using
singular value decomposition (SVD)

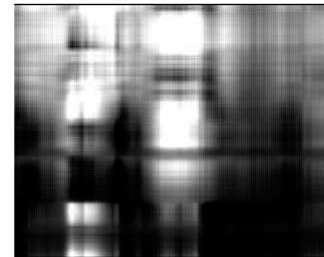
original



truncated SVD

less modes

$k = 3$



$k = 10$



$k = 20$



$k = 40$



more modes

$$\text{SVD: } M_{ab} = \sum_m u_{am} \Lambda_m (v^\dagger)_{mb} \approx \sum_{m=1}^k u_{am} \Lambda_m (v^\dagger)_{mb}$$

$$\Lambda_0 \geq \Lambda_1 \geq \dots$$

Application to $CP(1) + \theta$

2d CP(1) model

toy model of QCD

- asymptotic free
- confinement
- instanton

continuum

$$S = \int d^2x \left(\beta |D_\mu z|^2 + i \frac{\theta}{2\pi} \epsilon_{\mu\nu} \partial_\mu A_\nu \right)$$

U(1) auxiliary field

$$z \in \mathbb{C}^2$$

$$|z|^2 = \sum_{a=1}^2 z_a^* z_a = 1$$

on the lattice

$$S_{\text{lat}} = -2\beta \sum_{x,\mu} [z^\dagger(x) z(x + \hat{\mu}) U_\mu(x) + (c.c.)] - i \frac{\theta}{2\pi} \sum_x q(x) \in \mathbb{C}$$

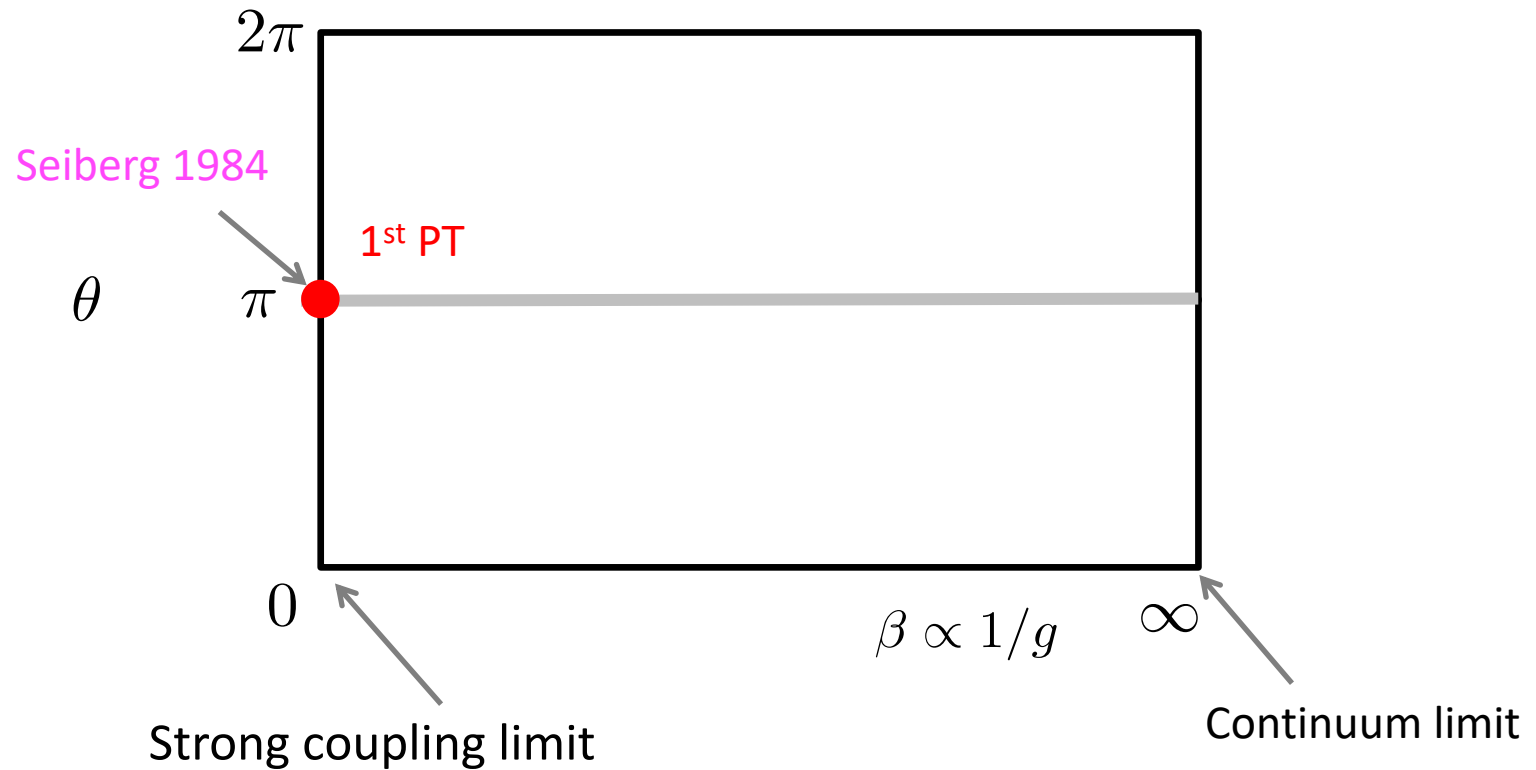
$$U_\mu = e^{iA_\mu}$$

sign problem!

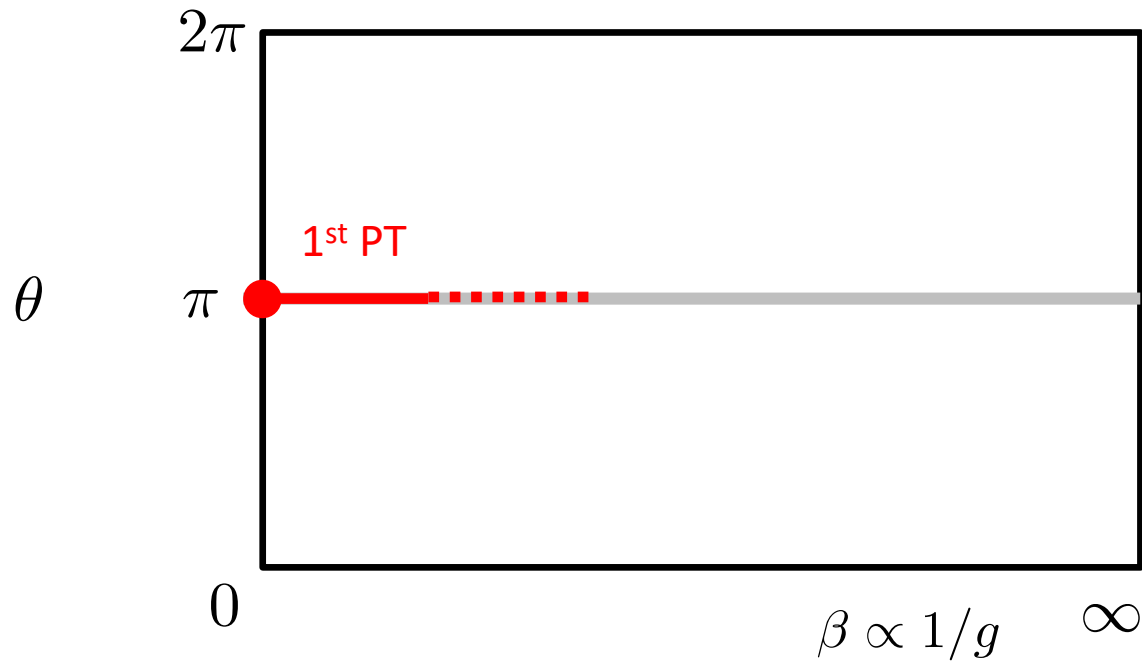


mod 2π

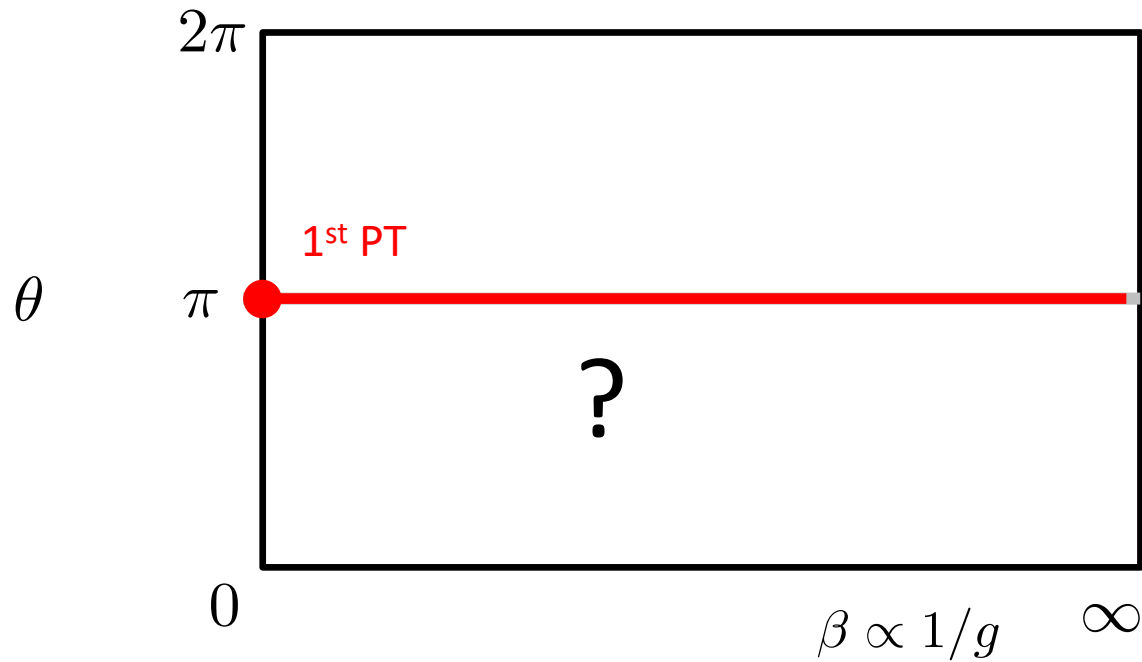
Previous study of phase structure



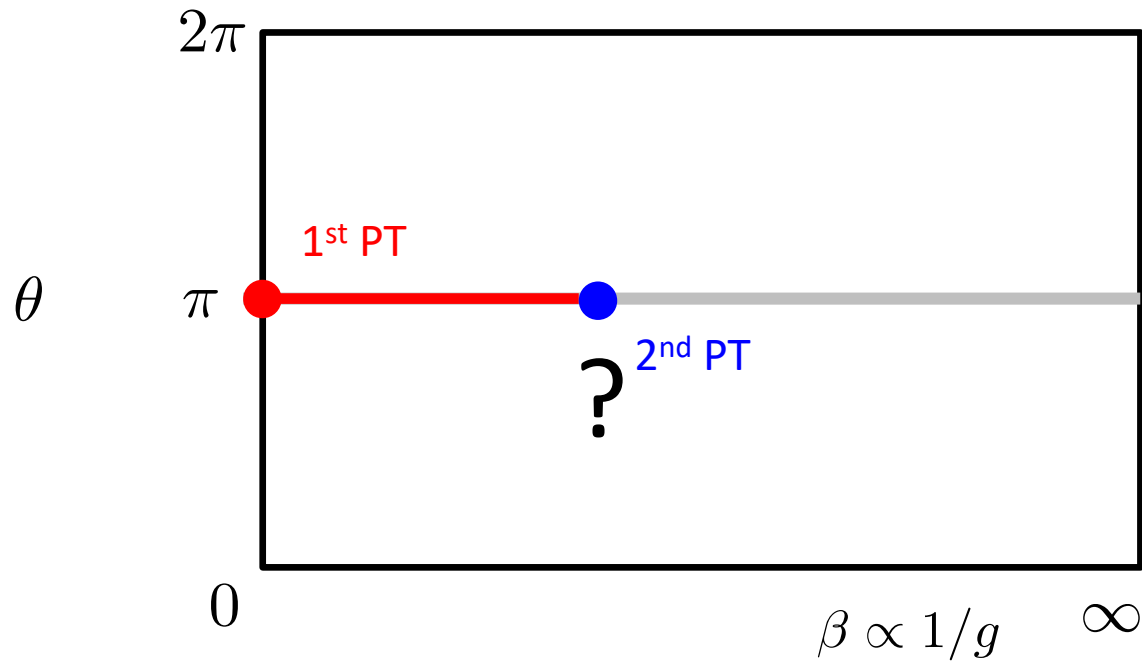
Previous study of phase structure



Previous study of phase structure



Previous study of phase structure

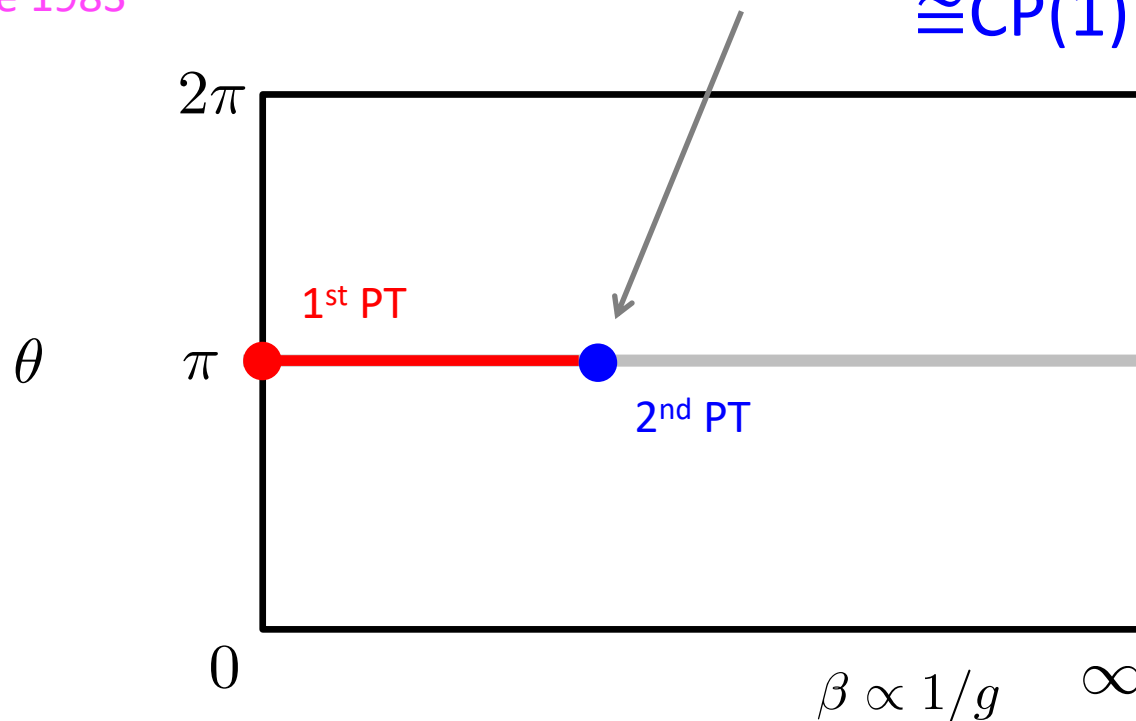


Previous study of phase structure

Haldane's conjecture : mass gap of $O(3)$ vanishes at $\theta=\pi$

Haldane 1983

$\cong CP(1)$



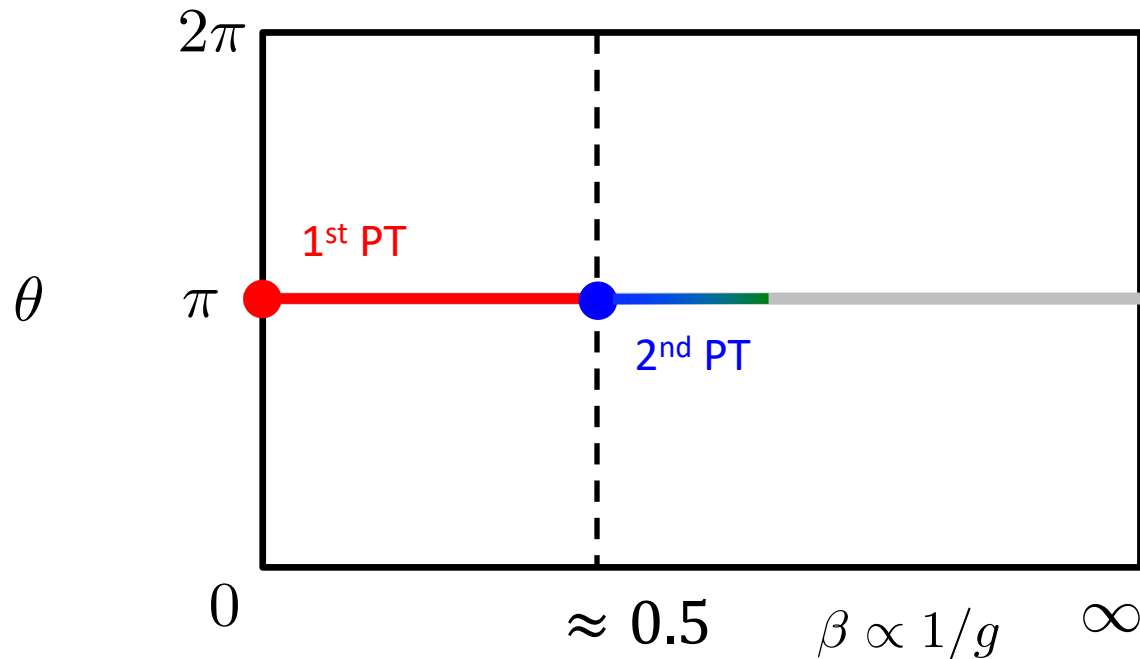
Affleck+Haldane 1987 expected universality class : $k = 1$ $SU(2)$ WZNW

Wess+Zumino 1971, Novikov 1981, Witten 1984

Previous study of phase structure

Azcoiti+ 2007

Monte Carlo, imaginary θ



Critical region is observed but the universality class is **not** identified

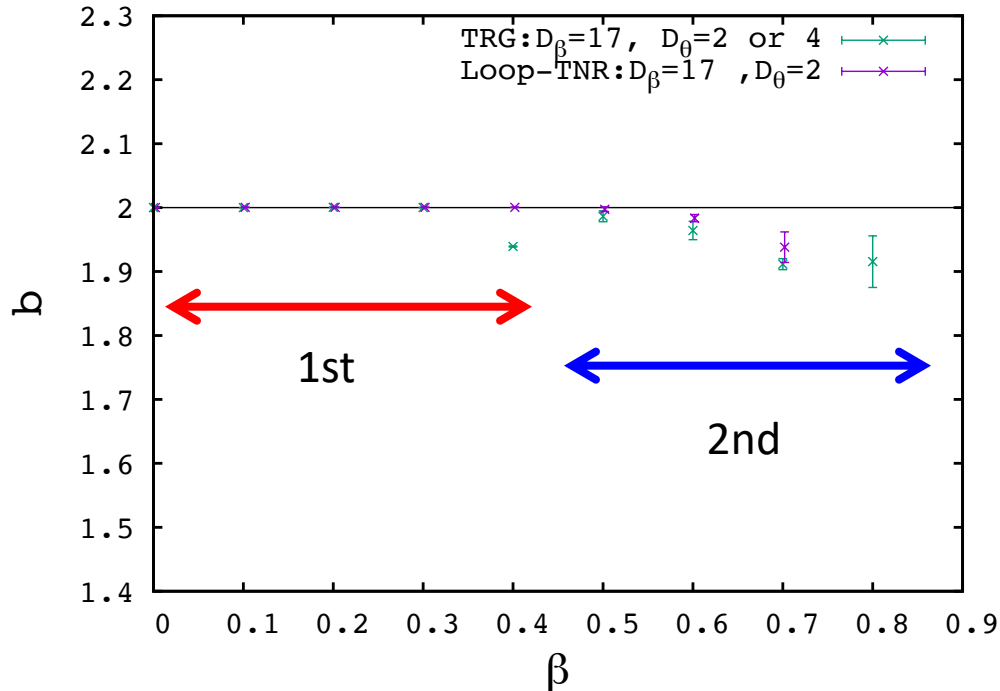
\Rightarrow This should be investigated by sign problem free method : **tensor network!**

Previous study of phase structure

Kawauchi+ST 2018

Tensor network

$$\chi = \frac{1}{L^2} \frac{\partial^2}{\partial \theta^2} \log \mathcal{Z}(\theta) |_{\theta=\pi} \propto L^b$$



1st : $b = 2$

2nd : $b = \gamma/\nu < 2$

$\beta = 0.5$ is roughly consistent with MC

Azcoiti+ 2007

But universality class is **not** determined

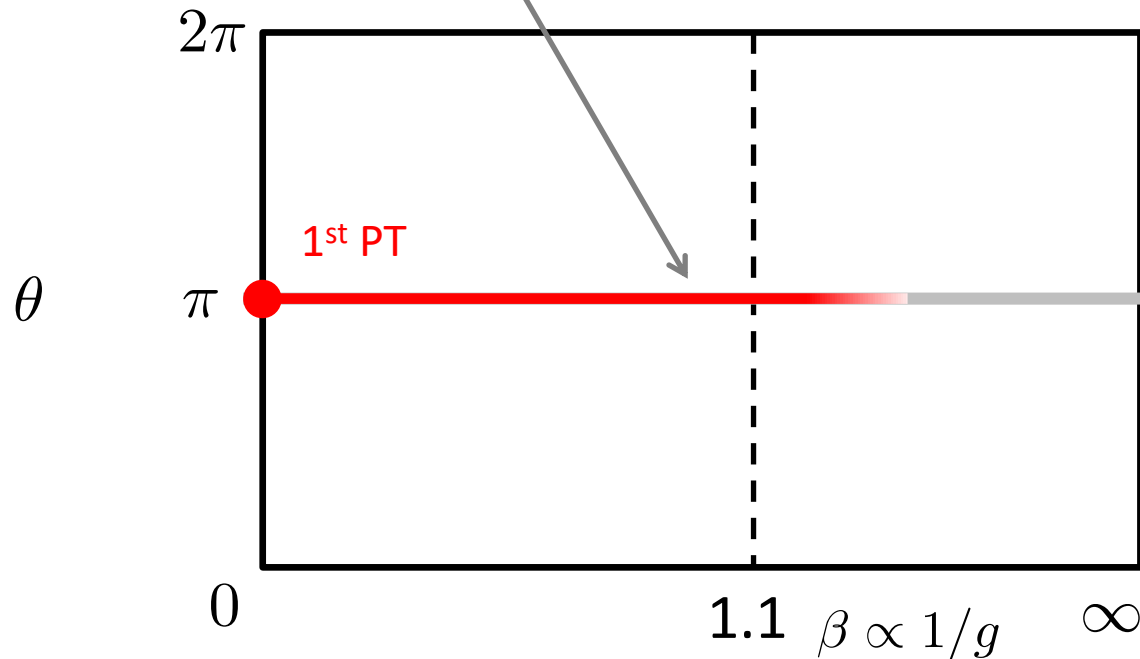
D_β, D_θ : truncation order

Previous study of phase structure

Nakayama+ 2021

Tensor network using Bond weighted TRG

Adachi+ 2020



no critical point is found up to $\beta=1.1$

What's new in our study

We make two improvements

- Using better initial tensor
- New analysis based on CFT

Improvement of initial tensor

Kawauchi+ST 2017, Nakayama+ 2021

- Previous studies use **character expansion for θ -term**

Hassan+ 1995

$$e^{i\frac{\theta}{2\pi}q_p} = \sum_{n \in \mathbb{Z}} e^{in(A_1 + A_2 - A_3 - A_4)} \underline{C_n(\theta)} \propto \frac{1}{n} \quad : \text{slow convergence!}$$

\Rightarrow large truncation error

- **New method** : Quadrature

Gauss-Legendre quadrature

$$\text{Gauge field : } \int_{-\pi}^{\pi} dA f(A) \approx \sum_{i=1}^{N_A} w_i f(A_i) \quad i \Rightarrow \text{tensor's index}$$

\uparrow
 weight

$$\text{Complex scalar field : } \int_{|z|^2=1} dz g(z) \approx \sum_{p=1}^{N_z} w_p^{(z)} g(z_p) \quad p \Rightarrow \text{tensor's index}$$

Genz+Keister 1996

Improvement of initial tensor

$$\delta f = \left| \frac{f_{\text{tensor}} - f_{\text{exact}}}{f_{\text{exact}}} \right|$$

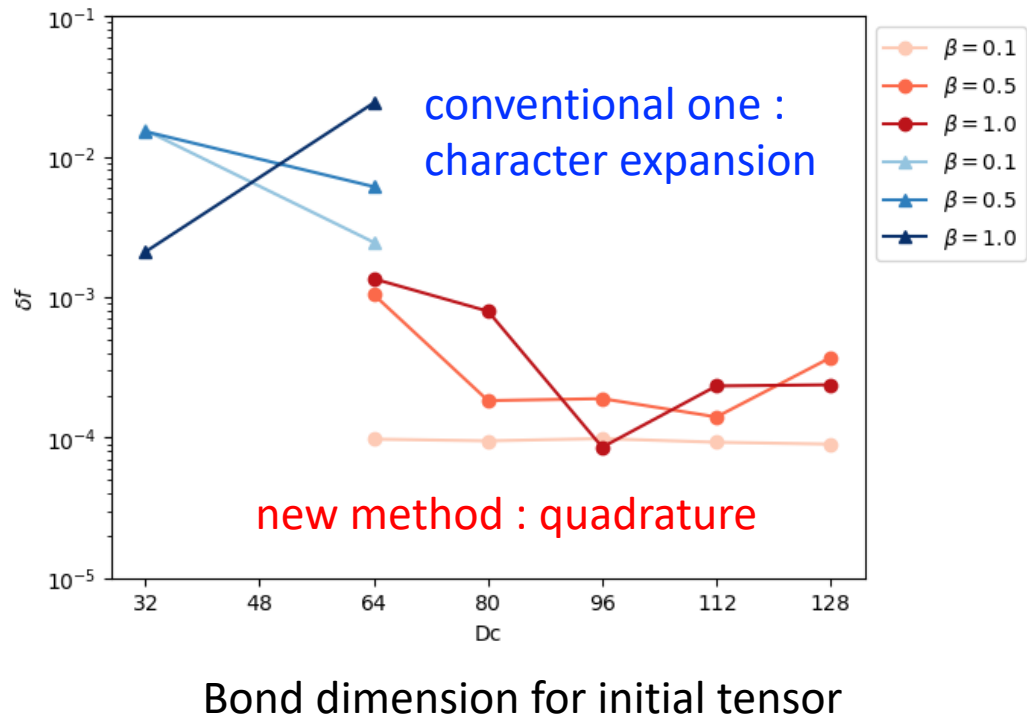
↓
free energy

$$V = 2 \times 2$$

$$\theta = \pi$$

$$N_A = 120$$

$$N_z = 224$$



New method shows better precision

New analysis method

- Previous studies use susceptibility

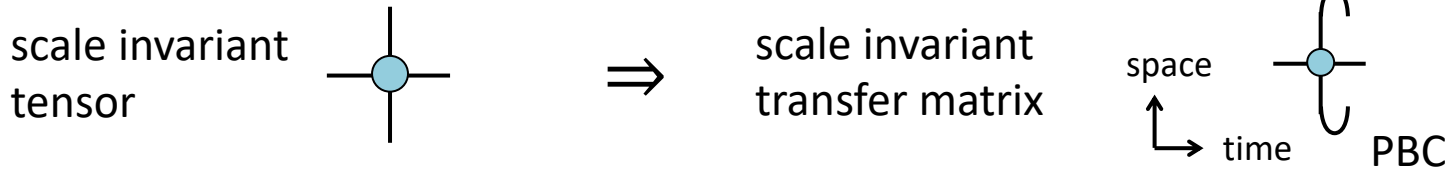
Kawauchi+ST 2017, Nakayama+ 2021

$$\chi = \frac{1}{L^2} \frac{\partial^2}{\partial \theta^2} \log \mathcal{Z}(\theta) |_{\theta=\pi} \propto L^b$$

fitting around $\theta = \pi$
 \Rightarrow fitting range ?

- CFT : central charge and scaling dimensions

Gu+Wen 2009



EV of TM

$$\Rightarrow \lambda_0 \geq \lambda_1 \geq \dots \Rightarrow$$

central charge

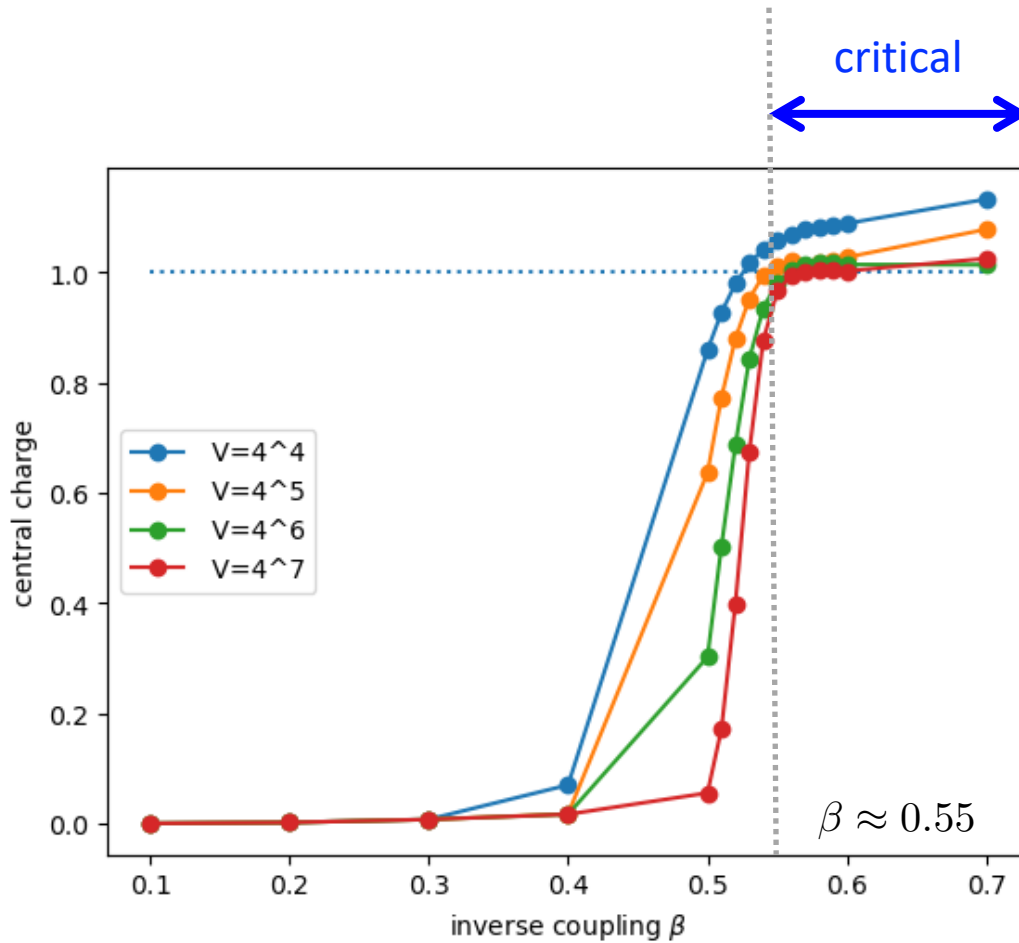
$$c = \frac{6}{\pi} \log(\lambda_0)$$

scaling dimension

$$x_i = h_i + \bar{h}_i = \frac{1}{2\pi} \log \left(\frac{\lambda_0}{\lambda_i} \right)$$

conformal weight

Central charge



critical $c = 1 \Leftrightarrow$ free boson CFT

$$\theta = \pi$$

$$N_A = 120$$

$$N_z = 224$$

$$\chi = 128$$

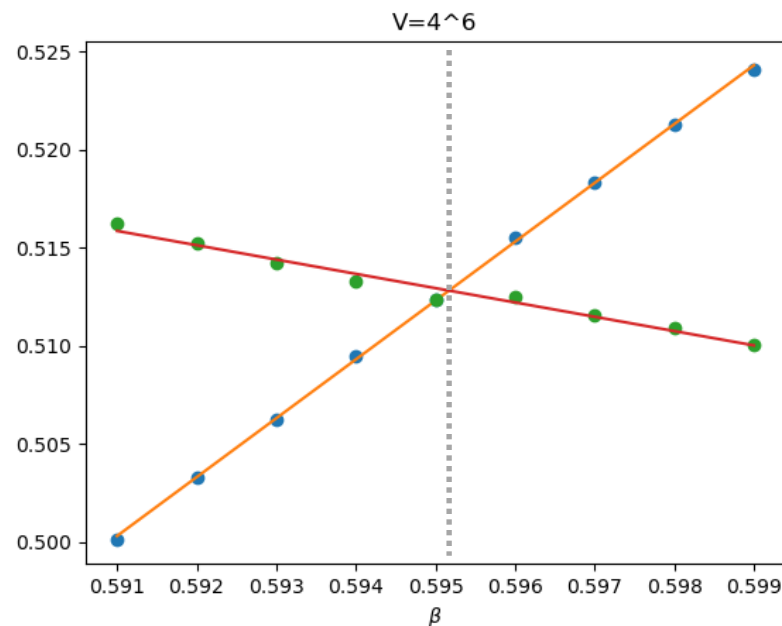
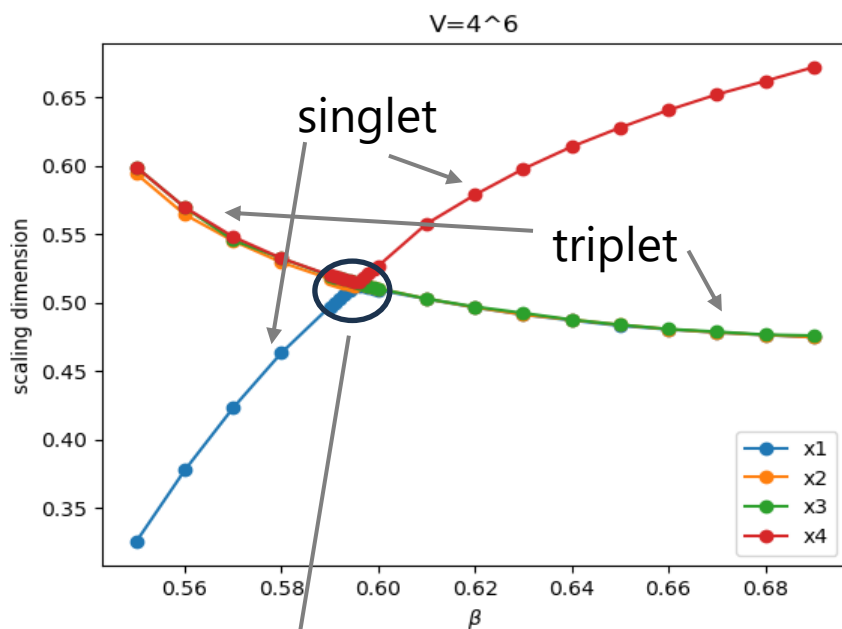
Bond weighted TRG

$$k = -0.5$$

Scaling dimensions

$$x_i = h_i + \bar{h}_i = \frac{1}{2\pi} \log \left(\frac{\lambda_0}{\lambda_i} \right)$$

$$x_1, x_2, x_3, x_4$$



$$\beta_{k=1SU(2)} \approx 0.595$$

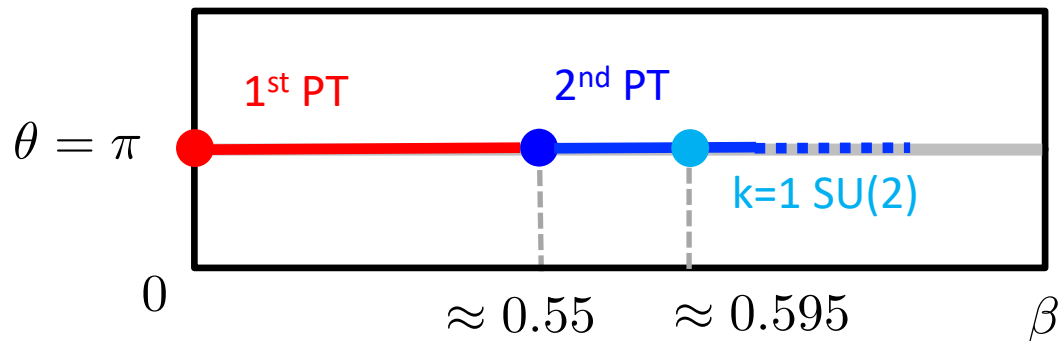
$$x_1 = x_2 = x_3 = x_4 \approx \frac{1}{2}$$

⇔ k=1 SU(2) WZNW

Wess+Zumino 1971, Novikov 1981, Witten 1984

Summary

- We have analyzed phase structure of 2d $CP(1)+\theta$ using tensor network
- **Two improvements**
 - Better (more precise) initial tensor
 - New analysis using CFT : central charge and scaling dimensions
- **Critical point region starts from $\beta \approx 0.55$ and $k=1$ $SU(2)$ symmetry is observed at $\beta \approx 0.595$**



German-Japanese Workshop 202? in Kanazawa

If our proposal is accepted!!!



I need your help!!!

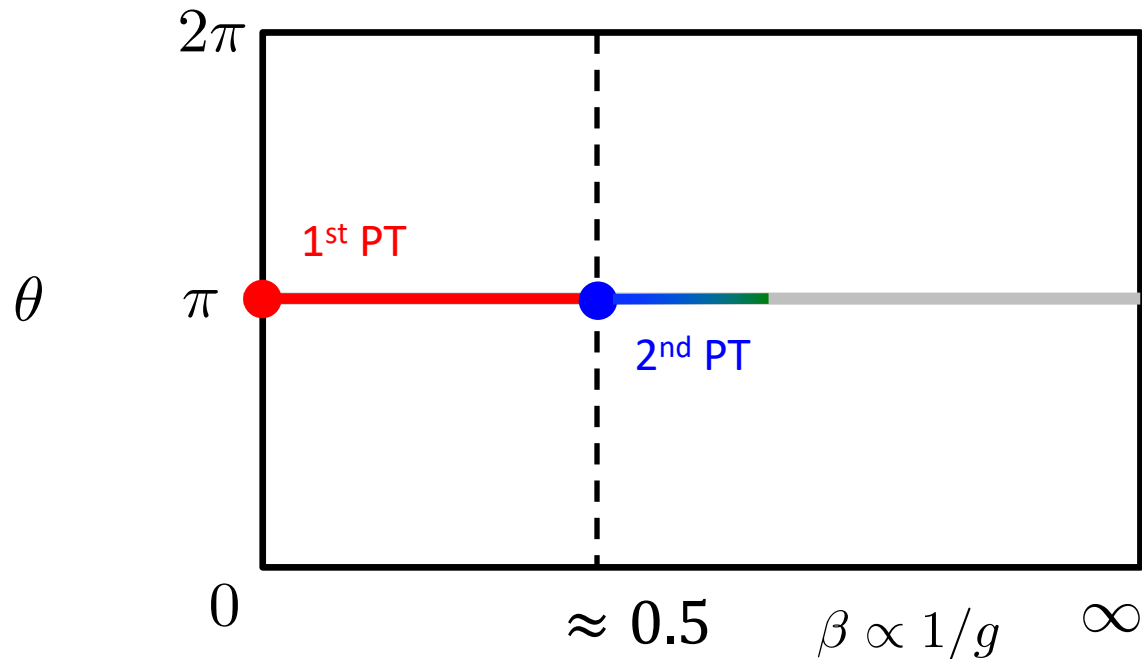
Future

- Mass gap scaling
- Continuum limit?

Back up

Previous study of phase structure

Kawauchi+ST 2018 Tensor network



Critical region is observed but the universality class is not identified

c=1, k=1 WZNW

scaling dimension for c=1 free boson CFT

$$x_i = x_{M,N} = \frac{M^2}{4R^2} + R^2 N^2, \quad M, N \in \mathbb{Z} \quad R : \text{compact radius}$$

for $R = \sqrt{\frac{1}{2}}$ \Rightarrow SU(2) symmetry appears \Rightarrow k=1 WZNW universality class
Ginsparg 1988

$$x_{\pm 1,0} = x_{0,\pm} = \frac{1}{2} \Rightarrow \text{quartet}$$

X

parameter

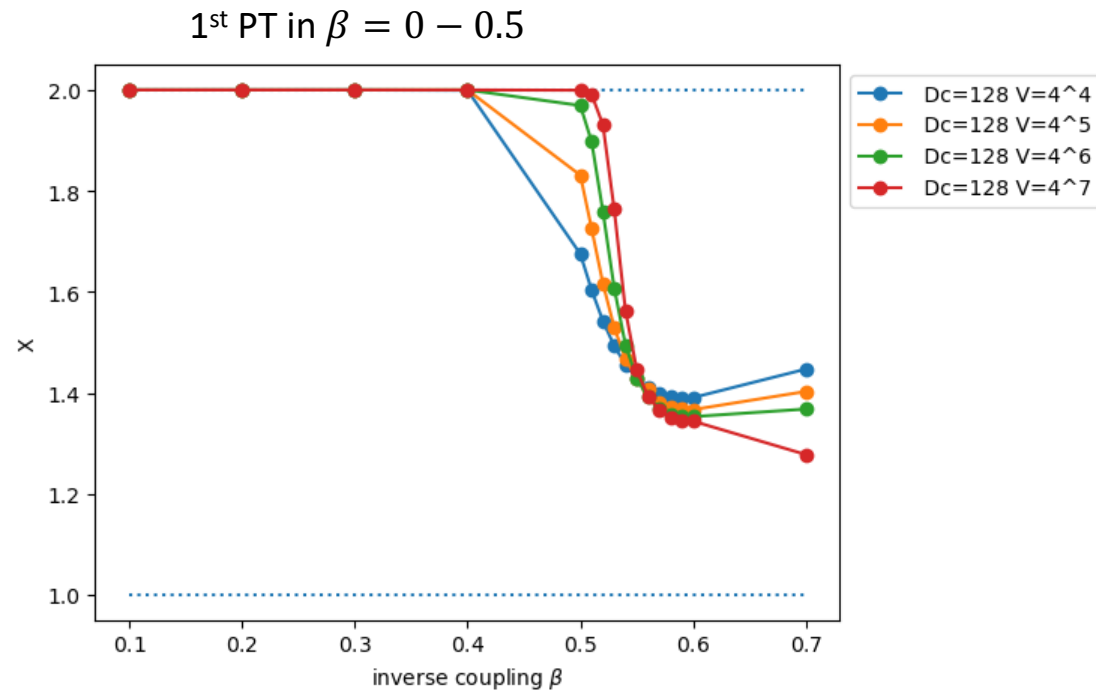
$$N_z = 226, N_A = 120, \\ Dc = 128, \theta = \pi$$

bond-weight TRG, $k = -1/2$

D. Adachi, T. Okubo, and S. Todo,
Phys. Rev. B 105, L060402(2022)

$$X = \frac{(\sum_i T_{ii})^2}{\sum_{ij} T_{ij} T_{ji}} = \frac{(\sum_i \exp[-E_i T])^2}{\sum_i \exp[-2E_i T]}$$

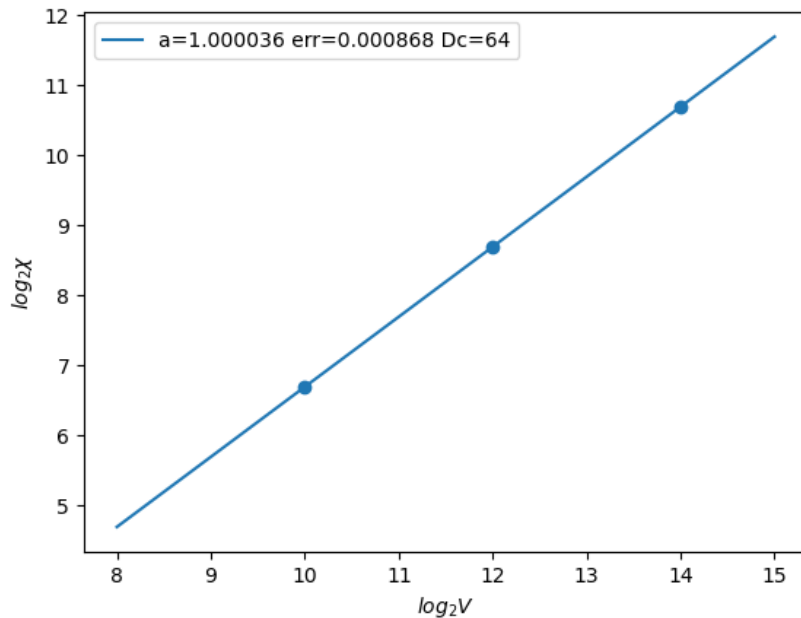
$$T \rightarrow \infty, \\ E_0 = E_1 \text{ のとき } X = 2$$



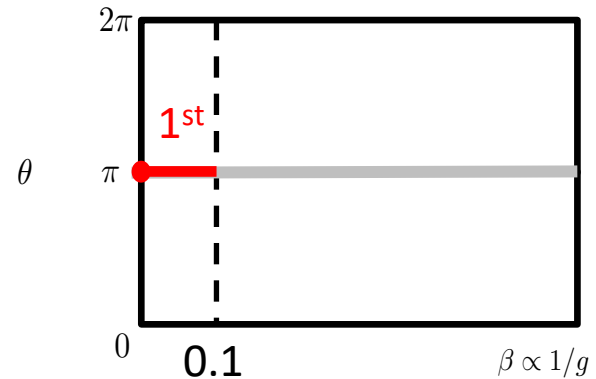
Nature of transition at $\beta = 0.1$

Kawauchi+ST 2017

$$\chi = \frac{1}{V} \frac{\partial^2}{\partial \theta^2} \log \mathcal{Z}(\theta) |_{\theta=\pi} \propto V^a \quad \begin{cases} 1\text{st} & : a = 1 \\ 2\text{nd} & : a = \gamma/\nu < 1 \end{cases}$$



For $\beta = 0.1$
 $a = 1.0000(9) \Rightarrow 1^{\text{st}} \text{ PT}$



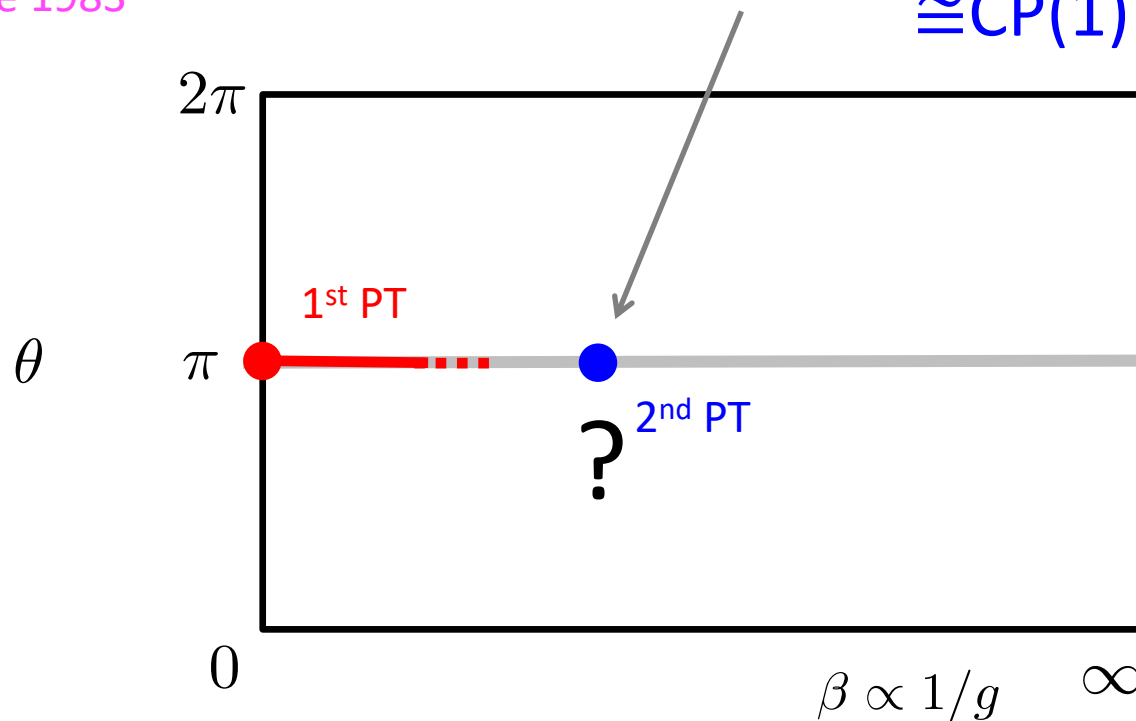
$D_{cut} = 64$, TRG for coarse-graining

Previous study of phase structure

Haldane's conjecture : mass gap of $O(3)$ vanishes at $\theta=\pi$

Haldane 1983

$\cong CP(1)$



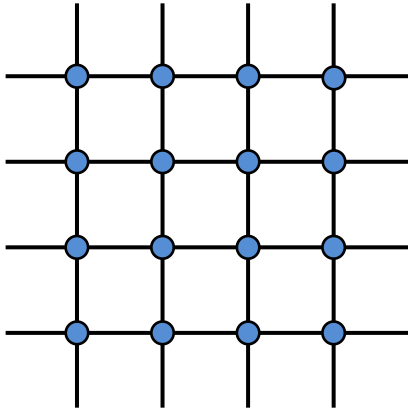
Bietenholtz et al. 1995, Wiese et al. 2012, de Forcrand et al. 2012, Azcoiti et al. 2012, Alles et al. 2014

$O(3) + \theta$ was intensively studied by MC and Haldane's conjecture is confirmed
And universality class is consistent with $k=1$ WZNW model

Wess & Zumino 1971, Novikov 1981, Witten 1984

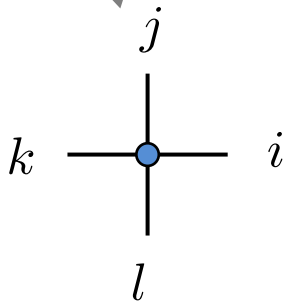
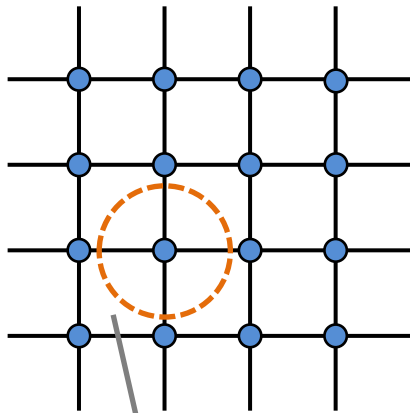
Coarse-graining

Tensor renormalization group (TRG)
[PRL99,120601\(2007\)](#)



Coarse-graining

Tensor renormalization group (TRG)
PRL99,120601(2007)

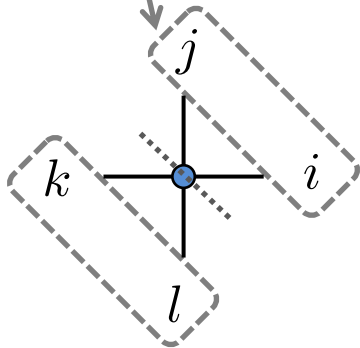
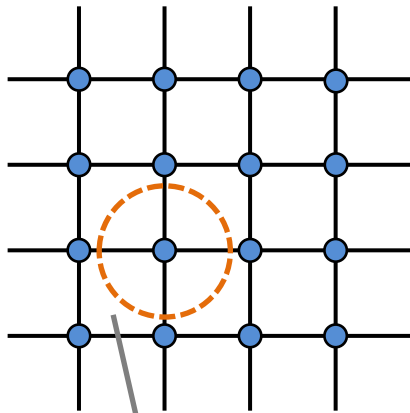


Bond dimension

$$1 \leq i, j, \dots \leq \chi$$
$$\Leftrightarrow T_{ijkl}$$

Coarse-graining

Tensor renormalization group (TRG)
PRL99,120601(2007)



$$1 \leq i, j, \dots \leq \chi$$

$$\Leftrightarrow T_{ijkl} = M_{(ij)(kl)}$$

$\chi^2 \times \chi^2$ matrix

Coarse-graining

Tensor renormalization group (TRG)
PRL99,120601(2007)

$$M \in \mathbb{C}^{x^2 \times x^2}$$

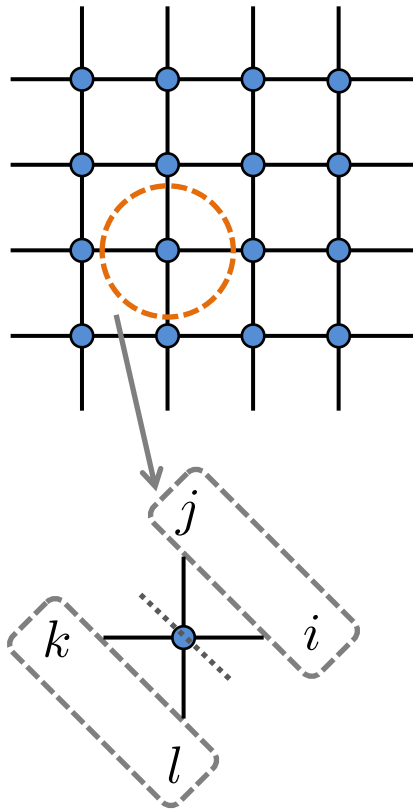
Singular Value Decomposition(SVD)

$$M_{ab} = \sum_m U_{am} \sigma_m (V^\dagger)_{mb}$$

unitary matrix

$$\sigma_1 \geq \sigma_2 \geq \dots \geq 0$$

: singular value (non-negative)



$$\Leftrightarrow T_{ijkl} = M_{(ij)(kl)}$$

Coarse-graining

Tensor renormalization group (TRG)
 PRL99,120601(2007)

$M \in \mathbb{C}^{\chi^2 \times \chi^2} \Rightarrow$ TN is sign-problem-free

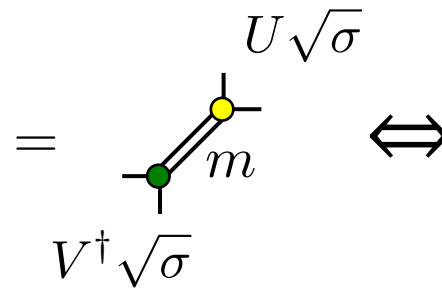
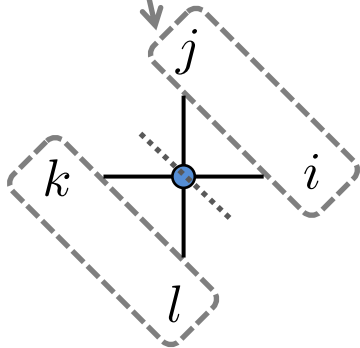
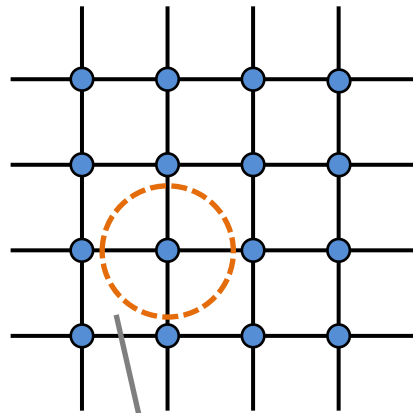
Singular Value Decomposition(SVD)

$$M_{ab} = \sum_m U_{am} \sigma_m (V^\dagger)_{mb}$$

\swarrow \nearrow
 unitary matrix

$$\sigma_1 \geq \sigma_2 \geq \dots \geq 0$$

: singular value (**non-negative**)



$$T_{ijkl} = M_{(ij)(kl)}$$

$$\stackrel{\text{SVD}}{=} \sum_{m=1}^{\chi^2} U_{(ij)m} \sigma_m V_{m(kl)}^\dagger$$

Coarse-graining

Tensor renormalization group (TRG)
 PRL99,120601(2007)

$M \in \mathbb{C}^{x^2 \times x^2} \Rightarrow$ TN is sign-problem-free

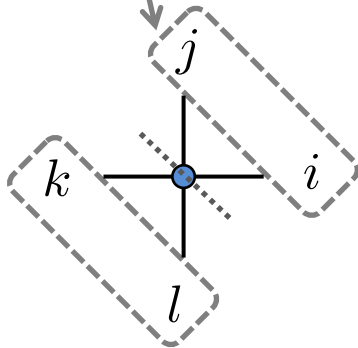
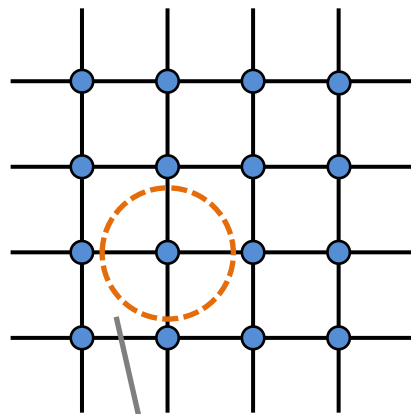
Singular Value Decomposition(SVD)

$$M_{ab} = \sum_m U_{am} \sigma_m (V^\dagger)_{mb}$$

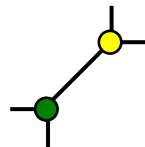
\swarrow \nearrow
 unitary matrix

$$\sigma_1 \geq \sigma_2 \geq \dots \geq 0$$

: singular value (**non-negative**)



\approx



\Leftrightarrow

$$T_{ijkl} = M_{(ij)(kl)}$$

truncation $\approx \sum_{m=1}^{\chi} U_{(ij)m} \sigma_m V_{m(kl)}^\dagger$

truncation of SVD = information compression

Tensor network rep. of Z

depends on property of field and interaction

- **Scalar field (non-compact)**
 - Orthonormal basis expansion
Shimizu mod.phys.lett. A27,1250035(2012), Lay & Rundnick PRL88,057203(2002)
 - Gauss Hermite quadrature *Sakai et al., JHEP03(2018)141*
- **Gauge field (compact : SU(N), CP(N) etc.)**
 - Character expansion : maintain symmetry, better convergence *Meurice et al., PRD88,056005(2013)*
- **Fermion field (Dirac/Majorana)**
 - Grassmann number $\theta^2=0 \rightarrow$ finite sum

In principle, we can treat any fields

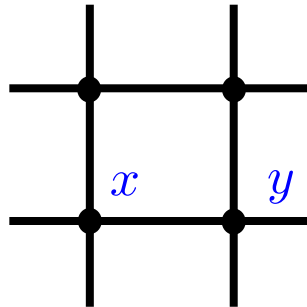
$$e^{\phi\theta} = 1 + \phi\theta = \sum_{n=0}^1 (\phi\theta)^n$$

Tensor network rep. for 2D Ising

e.g. 2D Ising model

$$Z = \sum_{\{s\}} \exp \left(\sum_{\langle x,y \rangle} \beta s_x s_y \right) = \sum_{\{s\}} \prod_{\langle x,y \rangle} \exp(\beta s_x s_y)$$

Nearest Neighbor



Tensor network rep. for 2D Ising

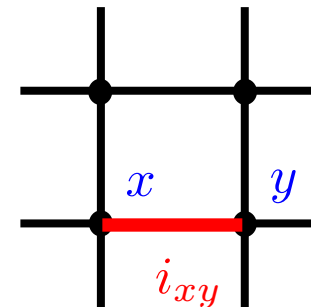
e.g. 2D Ising model

$$\begin{aligned} \mathcal{Z} &= \sum_{\{s\}} \exp\left(\sum_{\langle x,y \rangle} \beta s_x s_y\right) = \sum_{\{s\}} \prod_{\langle x,y \rangle} \exp(\beta s_x s_y) \\ &= (\cosh \beta)^{2V} \sum_{\{s\}} \prod_{\langle x,y \rangle} \sum_{i_{xy}=0}^1 (s_x s_y \tanh \beta)^{i_{xy}} \end{aligned}$$

$V = \#$ of lattice sites

$$\begin{aligned} \exp(\beta s_x s_y) &= \cosh(\beta s_x s_y) + \sinh(\beta s_x s_y) \\ &= \cosh \beta + s_x s_y \sinh \beta \\ &= \cosh \beta (1 + s_x s_y \tanh \beta) \\ &= \cosh \beta \sum_{i_{xy}=0}^1 (s_x s_y \tanh \beta)^{i_{xy}} \end{aligned}$$

$$s_x = \pm 1$$



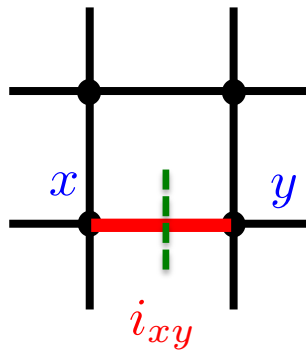
new d.o.f.

bond variable

Tensor network rep. for 2D Ising

e.g. 2D Ising model

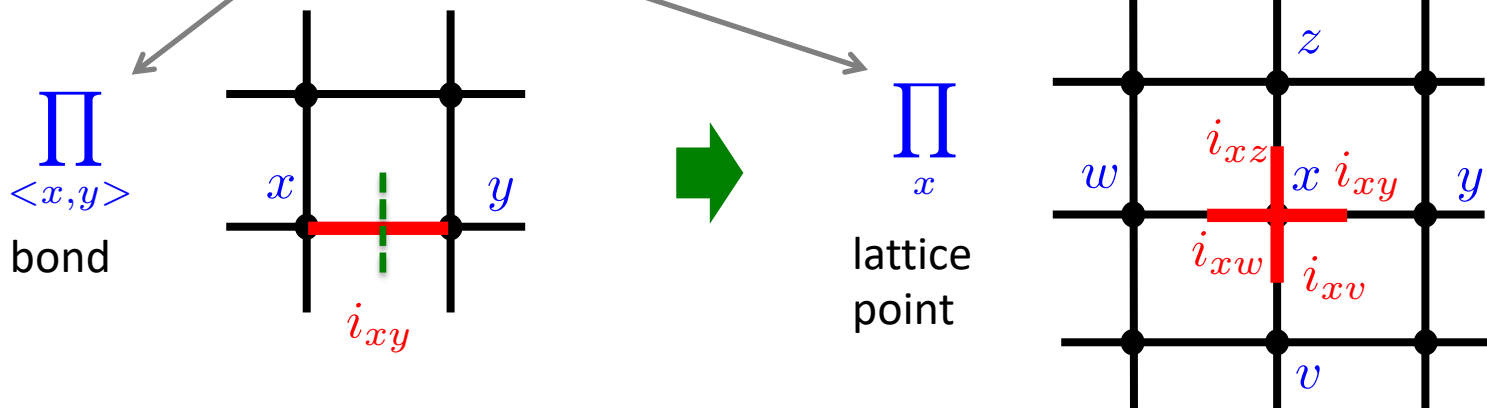
$$\begin{aligned} \mathcal{Z} &= \sum_{\{s\}} \exp \left(\sum_{\langle x,y \rangle} \beta s_x s_y \right) = \sum_{\{s\}} \prod_{\langle x,y \rangle} \exp(\beta s_x s_y) \\ &= (\cosh \beta)^{2V} \sum_{\{s\}} \prod_{\langle x,y \rangle} \sum_{i_{xy}=0}^1 (s_x s_y \tanh \beta)^{i_{xy}} \\ &= (\cosh \beta)^{2V} \sum_{\{i\}} \sum_{\{s\}} \prod_{\langle x,y \rangle} (s_x \sqrt{\tanh \beta} \cdot s_y \sqrt{\tanh \beta})^{i_{xy}} \end{aligned}$$



Tensor network rep. for 2D Ising

e.g. 2D Ising model

$$\begin{aligned}
 \mathcal{Z} &= \sum_{\{s\}} \exp \left(\sum_{\langle x,y \rangle} \beta s_x s_y \right) = \sum_{\{s\}} \prod_{\langle x,y \rangle} \exp(\beta s_x s_y) \\
 &= (\cosh \beta)^{2V} \sum_{\{s\}} \prod_{\langle x,y \rangle} \sum_{i_{xy}=0}^1 (s_x s_y \tanh \beta)^{i_{xy}} \\
 &= (\cosh \beta)^{2V} \sum_{\{i\}} \sum_{\{s\}} \prod_{\langle x,y \rangle} (s_x \sqrt{\tanh \beta} \cdot s_y \sqrt{\tanh \beta})^{i_{xy}} \\
 &= (\cosh \beta)^{2V} \sum_{\{i\}} \sum_{\{s\}} \prod_x (s_x \sqrt{\tanh \beta})^{i_{xy} + i_{xz} + i_{xw} + i_{xv}}
 \end{aligned}$$



Tensor network rep. for 2D Ising

e.g. 2D Ising model

$$\begin{aligned}
 \mathcal{Z} &= \sum_{\{s\}} \exp\left(\sum_{\langle x,y \rangle} \beta s_x s_y\right) = \sum_{\{s\}} \prod_{\langle x,y \rangle} \exp(\beta s_x s_y) \\
 &= (\cosh \beta)^{2V} \sum_{\{s\}} \prod_{\langle x,y \rangle} \sum_{i_{xy}=0}^1 (s_x s_y \tanh \beta)^{i_{xy}} \\
 &= (\cosh \beta)^{2V} \sum_{\{i\}} \sum_{\{s\}} \prod_{\langle x,y \rangle} (s_x \sqrt{\tanh \beta} \cdot s_y \sqrt{\tanh \beta})^{i_{xy}} \\
 &= (\cosh \beta)^{2V} \sum_{\{i\}} \sum_{\{s\}} \prod_x (s_x \sqrt{\tanh \beta})^{i_{xy} + i_{xz} + i_{xw} + i_{xv}} \\
 &= (\cosh \beta)^{2V} \sum_{\{i\}} \sum_{\{s\}} \prod_x (\sqrt{\tanh \beta})^{i_{xy} + i_{xz} + i_{xw} + i_{xv}} s_x^{i_{xy} + i_{xz} + i_{xw} + i_{xv}} \\
 &= (\cosh \beta)^{2V} \sum_{\{i\}} \prod_x (\sqrt{\tanh \beta})^{i_{xy} + i_{xz} + i_{xw} + i_{xv}} \sum_{s_x = \pm 1} s_x^{i_{xy} + i_{xz} + i_{xw} + i_{xv}} \quad \text{summation is done} \\
 &= (\cosh \beta)^{2V} \sum_{\{i\}} \prod_x (\sqrt{\tanh \beta})^{i_{xy} + i_{xz} + i_{xw} + i_{xv}} 2\delta(\text{mod}(i_{xy} + i_{xz} + i_{xw} + i_{xv}, 2)) \\
 &= T_{i_{xy} i_{xz} i_{xw} i_{xv}} \quad \text{new d.o.f. : index of tensor}
 \end{aligned}$$

Tensor network rep. for 2D Ising

$$\mathcal{Z} = 2^V (\cosh \beta)^{2V} \sum_{\dots, i, j, k, l, m, n, o, \dots} \cdots T_{ijkl} T_{mnop} \cdots$$

e.g. 2D Ising model

$$T_{ijkl} = (\sqrt{\tanh \beta})^{i+j+k+l} \delta(\text{mod}(i + j + k + l), 2)$$

$$\begin{bmatrix} T_{0000} & T_{0001} & T_{0010} & T_{0011} \\ T_{0100} & T_{0101} & T_{0110} & T_{0111} \\ T_{1000} & T_{1001} & T_{1010} & T_{1011} \\ T_{1100} & T_{1101} & T_{1110} & T_{1111} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & \tanh \beta \\ 0 & \tanh \beta & \tanh \beta & 0 \\ 0 & \tanh \beta & \tanh \beta & 0 \\ \tanh \beta & 0 & 0 & (\tanh \beta)^2 \end{bmatrix}$$

size and elements of tensor depend on a model

MC
Boltzmann weight is interpreted as probability
Importance sampling
Statistical errors
Sign problem may appear
Critical slowing down

TRG
Tensor network rep. of partition function (no probability interpretation)
Compression of tensor by SVD, Optimization
Systematic errors (truncated SVD)
No sign problem ∴ no probability
Efficiency of compression gets worse around criticality

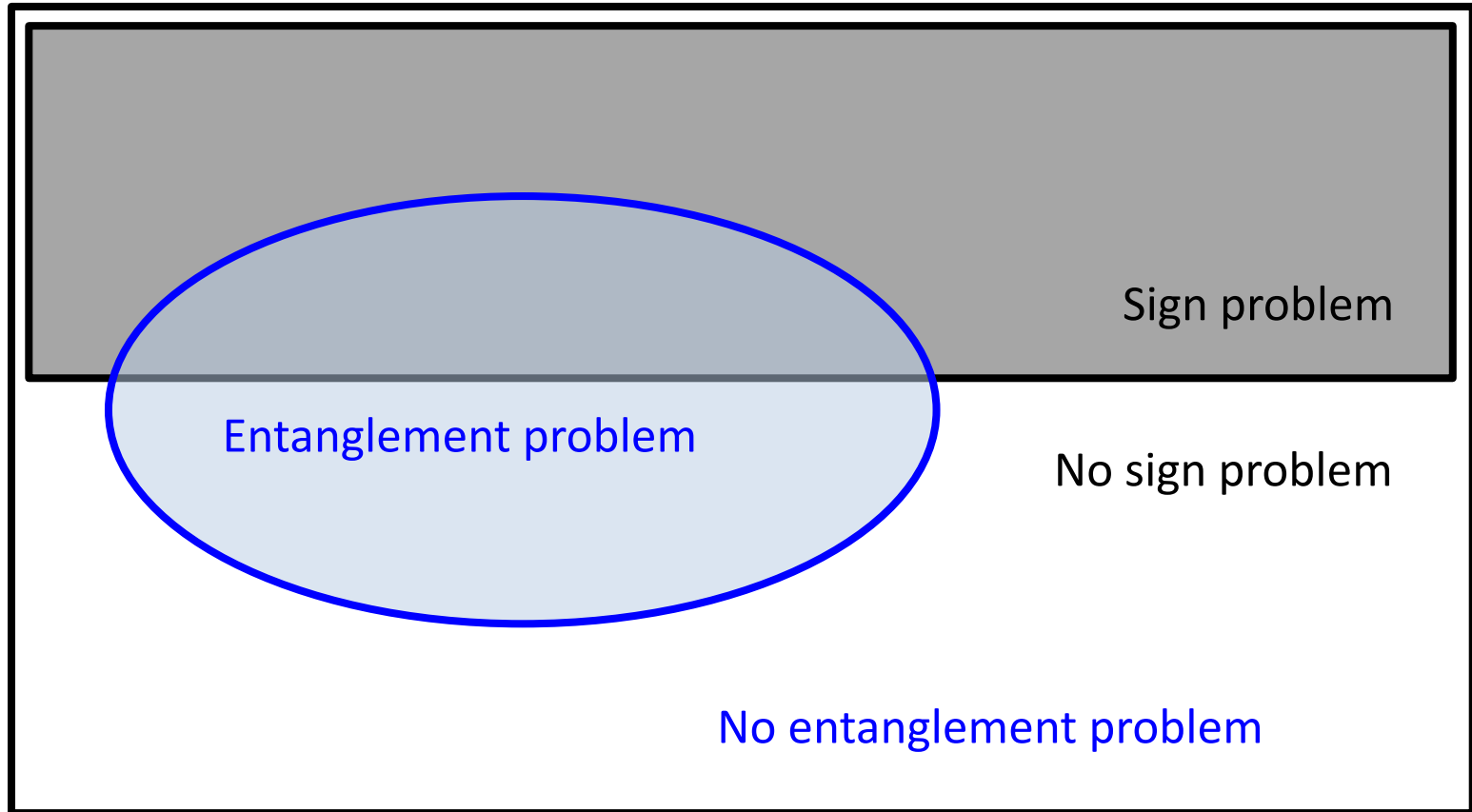


can be improved by TNR, Loop-TNR in 2D system
 Evenbly & Vidal 2014, Gu et al., 2015

From point of view of MC

Space of theory

From point of view of TN



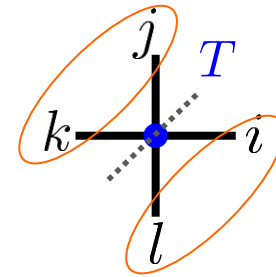
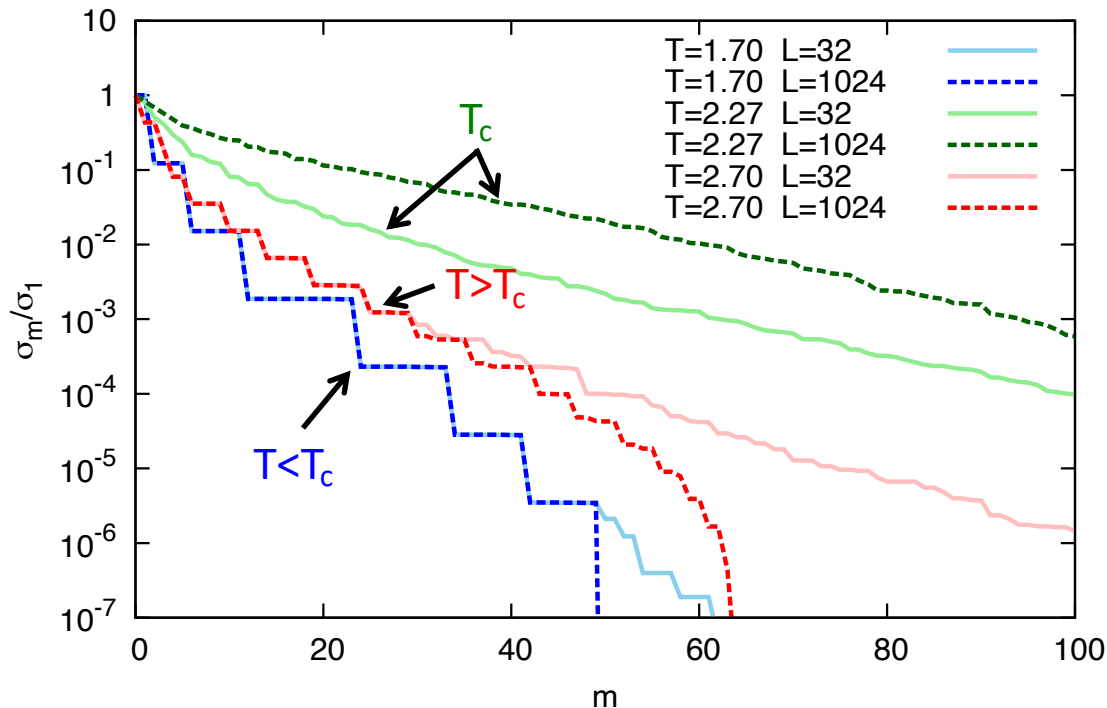
Is there a region where any method cannot access ?

Is entanglement problem NP-hard ?

Hierarchy of singular value

2D Ising model

$D_{\text{cut}}=32$



$$T_{(kj)(il)} = \sum_m u_{(kj)m} \sigma_m v_{m(il)}$$

$$L = 2^{\# \text{RG step}}$$

$$T_c = 2 / [\ln(1 + \sqrt{2})]$$

$$= 2.269\dots$$

- Off criticality: good hierarchy (small S)
- Near criticality: hierarchy gets worse (large S)

like critical slowing down in MC

Tensor network renormalization (TNR) [Evenbly&Vidal 2014](#) can help the situation

Renormalization group

$$H(K; \{s\}) = \sum_i K_i \mathcal{O}_i(\{s\})$$

s : spins

$$\mathcal{Z} = \sum_{\{s\}} e^{-\beta H}$$

Block spin transf., Migdal-Kadanoff RG

$$H = \sum_i K_i \mathcal{O}_i \xrightarrow{\mathcal{R}} H' = \sum_i K'_i \mathcal{O}'_i$$

$$K' = \mathcal{R}_K(K)$$

$$K^* = \mathcal{R}_K(K^*) : \text{fixed point}$$

target : critical exponent etc

TRG

$$T_{ijkl}(K)$$

i, j, k, l : indexes

$$\mathcal{Z} = \sum_{i,j,k,l,\dots} \prod T_{ijkl} \dots$$

SVD + contraction

$$\sum_{i,j,k,l,\dots} \prod T_{ijkl} \dots \xrightarrow{\mathcal{R}} \sum_{i,j,k,l,\dots} \prod T'_{ijkl} \dots$$

$$T' = \mathcal{R}_T(T)$$

$$T^* = \mathcal{R}_T(T^*) : \text{fixed point tensor}$$

target : partition function etc

SVD

rank- k $m \times n$ real matrix A ($m \geq n \geq k$) is given by

For simplicity
↙ ↘

$$A = U \Lambda V^T \quad (\text{full SVD})$$

$$\left\{ \begin{array}{l} U : m \times m \text{ orthonormal matrix : } U=(u_1, u_2, \dots, u_m), \quad U^T U = U U^T = I_m \\ V : n \times n \text{ orthonormal matrix : } V=(v_1, v_2, \dots, v_n), \quad V^T V = V V^T = I_n \\ \Lambda : m \times n \text{ diagonal matrix : } \Lambda = \text{diag}(\sigma_1, \sigma_2, \dots, \sigma_n) \end{array} \right.$$

where $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_k > \sigma_{k+1} = \dots = \sigma_n = 0$)

$$A = \sum_{l=1}^k \sigma_l u_l v_l^T \quad \text{Decomposition using rank-1 tensor}$$

Best approximation of A ?

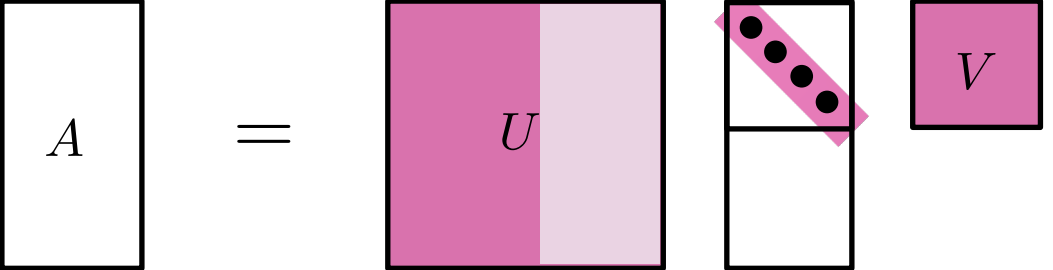
$$A = \sum_{l=1}^k \sigma_l u_l v_l^T$$


Diagram illustrating the decomposition of matrix A into its singular value decomposition: $A = U \Sigma V^T$. Matrix A is shown as a vertical rectangle. It is equal to matrix U (a square with a pink left half and a light pink right half), matrix Σ (a vertical rectangle with a pink top half containing four black dots on a diagonal line and a white bottom half), and matrix V (a square with a pink top half and a white bottom half).

truncated at r

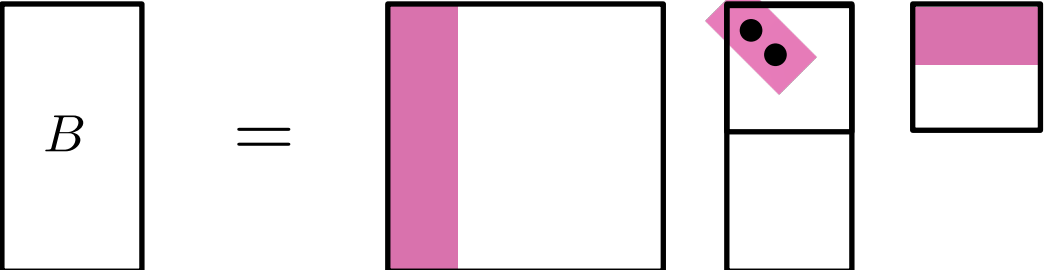
$$B = \sum_{l=1}^r \sigma_l u_l v_l^T$$


Diagram illustrating the truncated singular value decomposition of matrix A to form matrix B . Matrix B is shown as a vertical rectangle. It is equal to matrix U (a square with a pink left half and a white right half), matrix Σ (a vertical rectangle with a pink top half containing two black dots on a diagonal line and a white bottom half), and matrix V (a square with a pink top half and a white bottom half). A green arrow points from the Σ matrix in the top diagram to the Σ matrix in this diagram, indicating the truncation process.