國立台灣大學理學院物理學系

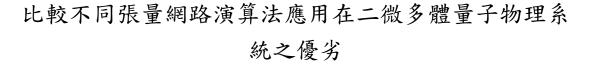
碩士論文

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Comparison between Tensor Network Algorithms for Two Dimensional Infinite Quantum Many-Body Systems

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隨著時光流逝,終於完成了我在物理系的碩班生涯。比起大學在材料系的無聊日子,這三年的經歷讓我感到彌足珍貴。

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摘要

如何判斷多體量子系統的相變化,且從微觀系統來得到巨觀上的物理性質,在現代仍為凝態物理學中十分有趣的領域。

從NRG的為起點開始,多年來出現了許多突破性的演算法。其中 DMRG在一維的系統的模擬中得到了相當好的結果。但在二微系統 中,因為 Area law 的關係使其表現不如在一微系統中精確,不僅如此, 在二維系統中,計算複雜度上升之速度也非一維系統可比擬。為了解 決這些問題,因而出現了許許多多不同的建立在張亮網路理論上的演 算法。

此篇論文,紀錄了幾個當今較為主流或新穎並用以模疑二維量子系統的張亮網路演算法。一開始將簡單解釋張亮網路的基本理論;再來會介紹如何實做、優化演算法,以增加精確度和降低計算複雜度。章節中也附上偽代碼,來說明實作中應注意之細節。最後會比較它們計算二維易辛模型與海森堡模型的結果,來說明各演算法之優缺點。



Abstract

Determining the phase transition of many body systems and the physical properties of macroscopic systems from microscopic description are still challenging in condense matter physics.

Since the numerical renormalization group (NRG) came out, various algorithms sprang up like mushrooms for analyzing these problem. Among all, the density matrix renormalization group (DMRG) could be considered as the most remarkable outcome, which analyze accurately in one dimensional systems. However, it perform worse in two dimensional systems. Not only the physical reasons, such as the area law, but also the rapid increment of computational complexity which is much higher than one dimensional system.

In the thesis, we recorded some of popular tensor network algorithms which are developed for handling the problems in two-dimensional systems. First of all, we briefly introduce the tensor network theory. Then, in the following sections, we shown the network diagrams and simple pseudocode for explaining how to implement these algorithms.

Key words— matrix product state(MPS), projected entangled pair state(PEPS), projected entangled simplex state, infinite time-evolveing block-decimation, corner transfer matrix, tensor renormalization group, Benchmarks, uni10.



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Chapter 1

Introduction

1.1 Overview

Understanding the phenomena of quantum many-body systems is one of most challenging problem in condense matter physics. Since, the coefficients required for describing entire systems grow exponentially with system size. For instance, If we desired to fully describe a N-site spin chain and each spin has d probable states, the requirement of coefficients is d^N . Due to the rapid increment of computational consumptions, it's impossible to simulate the system whose size is larger than 50 for classical computers.

For addressing that problems, various numerical methods have been developed. For instance, density matrix renormalization group (DMRG) [1] [2] acquire the accuracy ground state energy and provide a dominant tool to study properties of one dimensional systems. Furthermore, the theory of DMRG is related with matrix product states (MPS) [3] [4], which could describe the wave-function of one dimensional system and be explicitly represented by *tensor diagrams*.

No doubt, DMRG is the most successful method in one-dimensional systems. However, it's failure in higher dimensional systems due to the insufficiency of dealing with the entanglement in systems from matrix product states. For two-dimensional lattices, the projected entangled pair state (PEPS) [5] [6] has been applied to deal with that problem and many algorithms sprang up like mushrooms, such as time-evolving block decimation [7] [8] and multiscale entanglement renormalization ansatz [9].



Chapter 2

Tensor Network Theory

This section begins from a fundamental question: How to draw a tensor network diagrams? In tensor network theory [10] [11] [12], we used to represent tensors graphically instead of complicated equations, because *tensor diagrams* can map to quantum states and geometric lattices explicitly. Base on its clear representation, the implementation of tensor network algorithms become simply.

2.1 Representation of tensors in tensor Networks

In mathematical concept, a tensor is considered as a multi-dimensional array of scalers. The arrangement of the elements in a tensor is dependent on its *indices* and the *rank* of tensor is equivalent to the number of indices. Thus, a rank-0 tensor is a scaler (T), a rank-1 tensor is a vector (T_i) , a rank-2 tensor is a matrix (T_{ij}) and so on.

Graphically, we usually use a node and few bonds to represent a tensor. Some specified examples are shown in figure 2.1, the number of bond is equal to the rank of tensor, which means that tensors without bonds, with a single bond, with two bonds and with three bonds can map to scalers, vectors, matrices and rank-3 tensor.

Then, we should determine the dimension of tensors. Explicitly, the dimension of

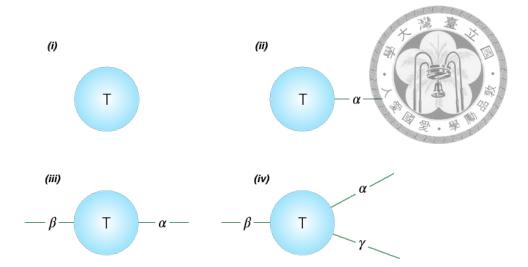


Figure 2.1: Usually we use a note and few bonds to compose a tensor and the numbers of bond depend on the rank of tensor. (i) A tensor without bonds is a scaler T, (ii) A tensor with one bond is a vector T_{α} , (iii) A tensor with two bonds is a Matrix $T_{\alpha\beta}$, (iv) A tensor with three bonds is a rank-3 tensor $T_{\alpha\beta\gamma}$.

rank=0 (scaler) is equal to one. However, the dimension of tensors higher than rank-0 depend on the bond dimensions. For instance, in Fig.2.1(iv), it's a rank-3 tensor and the dimensions of the indices are χ_{α} , χ_{β} , χ_{γ} and T contains $\chi_{\alpha}\chi_{\beta}\chi_{\gamma}$ components.

2.2 Tensor operations and tensor network diagrams

Basically we can't calculate an fold tensors directly. So the first step, we should unfold tensors. On the other words, we must make their rank lower than 3 before operating. The process is also called *permuteation*.

2.2.1 Permutation

Fig2.2 is a simple permutation example which means that the tensor \hat{A} permuted into tensor \hat{A} ,

$$A_{\alpha\beta\gamma} \to \hat{A}_{\alpha\gamma\beta}$$
 (2.1)

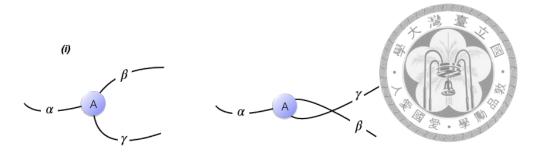


Figure 2.2: Simple example of a tensor permutation. Permute tensor A from indices $\alpha\beta\gamma$ to $\hat{A}_{\alpha\gamma\beta}$

The components of \hat{A} and A are equivalence, but having different arrangements. In order to explaining clearly, I use two flags, incoming (BD_IN) and outgoing (BD_OUT) bonds which are also designed for distinguishing different types of uni10::Bond in Uni10, to show how to reduce the rank of a tensor. Further explanation, BD_IN and BD_OUT can be imagined as rows and columns of a matrix. For instance, if the indices of $T_{\alpha\beta\gamma}$ ordered like Fig2.3(i), it means that $T_{\alpha\beta\gamma}$ is a matrix $T_{\chi_{\beta},\chi_{\alpha},\chi_{\alpha}}$. Similarity, If it's like Fig2.3(ii), it means that $T_{\alpha\beta\gamma}$ is a matrix $T_{\chi_{\beta},\chi_{\alpha},\chi_{\alpha}}$.

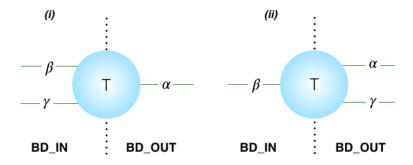


Figure 2.3: (i) Unfold a tensor to a matrix $T_{\chi_{\beta}\chi_{\gamma},\chi_{\alpha}}$, (ii) Unfold a tensor to a matrix $T_{\chi_{\beta},\chi_{\alpha}\chi_{\gamma}}$.

2.2.2 Tensor contraction

Tensor contraction is defined as the sum of all products of the shared indices of tensors. For instance, tensor contraction between two rank-2 tensors $A_{\alpha\beta}$ and $B_{\beta\gamma}$ which is equivalent to inner product between matrix, inner product between matrix $A_{\chi\alpha,\chi\beta}$ and

$$B_{\chi_{\beta},\chi_{\gamma}}$$

$$C_{\alpha\gamma} = \sum_{\beta=1}^{\chi_{\beta}} A_{\alpha\beta} B_{\beta\gamma},$$



and the tensor diagram is shown in Fig2.4(i),

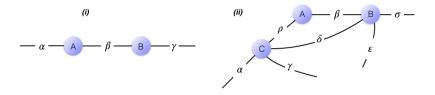


Figure 2.4: (i) Contraction between $A_{\alpha\beta}$ and $B_{\beta\gamma}$ is equivalent to inner product between matrix $A_{\chi_{\alpha},\chi_{\beta}}$ and $B_{\chi_{\beta},\chi_{\gamma}}$ (ii) A simple tensor network

Now we considered a more complexity network, in Fig2.4(ii). The equations is written as,

$$D_{\alpha\gamma\sigma\epsilon} = \sum_{\beta\rho\delta} A_{\rho\beta} B_{\beta\sigma\epsilon\delta} C_{\gamma\delta\rho\alpha}, \qquad (2.3)$$

The contraction processes of this network can be separated to some steps.

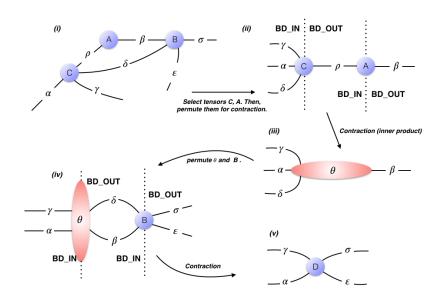


Figure 2.5: The contraction processes of the example shown in Fig2.4

(i) First, select a pair of tensors arbitrarily. In Fig2.5, we chose beginning from contracting tensors A and C . Actually, this step is very crucial. For example, If the dimensions of each bonds are the same and we select tensor B and C instead, it not only enlarge computational consumption, but also make algorithms inefficient. I'll discuss more detail in $\ref{eq:constraint}$ (ii) Second, A and C should permute into legal shape. In this case, A and C are permuted into $A_{\gamma\alpha\delta\rho}$ with one BD_OUT and $C_{\rho\beta}$ with one BD_IN, therefore the question evolves to inner product matrix $A_{\chi\gamma\chi\alpha\chi\delta,\chi_\rho}$ and A_{χ_ρ,χ_β} . Repeated step (i) and (ii) again, after getting the tensor $\theta_{\gamma\alpha\delta\beta}$. Permute tensors θ and B into $\theta_{\gamma\alpha\delta\beta}$ having two BD_OUT and $B_{\delta\beta\sigma\epsilon}$ having two BD_IN and we will get tensor D, after contracting them by matrix multiplication.

2.3 Describe Quantum states with tensor network

Before drawing a many-body system with TN, we should discuss how to describe a chain of N particles, with each particle having d different states and why using TN. Hence, we considered a many-body systems is composed of many local particles and have studied that a pure state corresponds to a vector in Hilbert space, so we intuitively imagined that the wave-function of systems can be written as,

$$|\Psi_N\rangle = \sum_{i_1, i_2, \dots, i_N} C_{i_1, i_2, i_3, \dots, i_N} |i_1\rangle \otimes |i_2\rangle \otimes \dots \otimes |i_N\rangle,$$
 (2.4)

each individual $|i_1\rangle, |i_2\rangle, \dots, |i_N\rangle$ spanned by an orthogonal basis which the degree of freedom is d. After writing down the eq.2.4, we are able to build a TN for quantum states, each bond of tensor corresponds to the local Hilbert space and the dimensions of each bond is equivalent to the states of particle on i site. The geometric notation of $|\Psi\rangle$ and some fundamental operations are shown in Fig.2.6.

From 2.4, we are aware of the number of coefficients, $C_{i_1,i_2,i_3,\dots,N}$, is proportional to d^N and fully describing a many-body system which is too expensive. Fortunately, the wavefunction can be decomposed to two subsystem by Schmidt Decomposition and there are

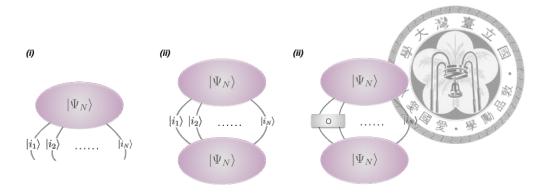


Figure 2.6: (i) Wave-function: $|\Psi_N\rangle$ (ii) Norm of $|\Psi_N\rangle$; $\langle\Psi_N | \Psi_N\rangle$ (iii) Expectation value of observable O; $\langle\Psi_N | O | \Psi_N\rangle$

more details in chapter.3.



Chapter 3

2-D Imaginary Time Evolving Block

Decimation

In this chapter we introduce some different ways to implement two-dimensional imaginary time-evolving block decimation and apply them in calculation of ground state of Heisenberg and transverse Ising model on two-dimensional square lattice. In section 3.1 and 3.2, we briefly review the idea of imaginary time evolution (TEBD) [7] [8] and explain how to extend it to an infinite two-dimensional system [13]. Second, we briefly review another implementation which is able to make 2D-iTEBD more stable[14]. In the last section 3.4, we record various particulars which are helpful optimizing algorithms.

3.1 Imaginary Time Evolution

Theoretically, if having the imaginary time evolution operator $e^{-\tau H}$, we could project any random states to the ground state, as long as the wave-function can be written as,

$$|\psi_0\rangle = \frac{e^{\tau H} |\Psi\rangle}{\parallel e^{\tau H} |\Psi\rangle \parallel} \tag{3.1}$$

but according to the eq.3.1, we may found that the number of coefficients in an origin evolution operator $e^{-\tau H}$ is proportional to $2^N \times 2^N$. On the other words, it is impossible

to update entire system directly. In order to restricting the rapid dimensional growth, we apply *Suzuki Trotter decomposition* to approximate. The main idea of *Suzuki Trotter* is decomposing the whole system to lots of units cell and using some smaller operations to update the wave-function.

$$e^{\delta A+B} = e^{\delta A}e^{\delta B} + O(\delta^2) \tag{3.2}$$

eq.3.2 means the first-order Suzuki Trotter decomposition, and A and B are non-commutative with each other.

Now that the dimension of the evolution operator is reduced to a n-site operator and in chapter ??, we have shown that a many-body system can map to a MPS [3][4] or PEPS [5] structure, so we can draw the process of updating a ground state like Fig.3.1,

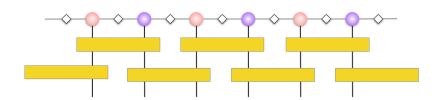


Figure 3.1: The red and blue tensor denotes on *odd* and *even* sites. The yellow one are time evolution operators $e^{-\tau H_{k,k+1}}$, $e^{-\tau H_{k+1,k}}$

On the other work, contract the tensors in Fig.3.1 repeatedly until the ground state energy to the minimum. The remained tensor is considered as the ground state of the system. So the next question: How can we contract them and preserve the structure like Fig3.2? This answer is iTEBD.

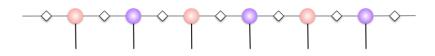


Figure 3.2: The simple form of a matrix product state.

3.2 Simple Infinite Imaginary-Time Evolving Block Dec-

imation for 2-D system

In this section, we apply the TN diagrams to introduce various implementation methods. More theoretical details are included in the references [15] [16] [17].

3.2.1 Simple Description of iTEBD for 1-D system

The algorithm start from 2 random states and 2 random diagonal matrices which are considered as entanglement between particles. In the TN diagrams, Fig.3.3, the states and entanglement between neighbor sites are represented by the nodes and bonds with different colors.

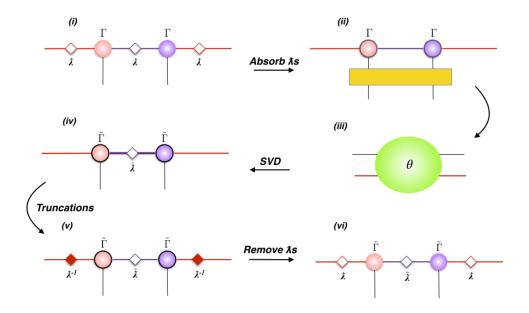


Figure 3.3: (i)Absorb all λ to Γ . (ii) Contract an evolution operator $e^{-\delta H}$ for evolving the system. (iii) Decompose the tensor θ by SVD. (iv) Truncate and Update the states and λ on the green bond.(v) Remove λ for obtaining the states. (iv) After updating the states and λ on the purple bond, apply the way to update the red bond and repeat all the steps until the ground state convergence.

The processes shown in Fig.3.3 is a standard strategy for implementation of iTEBD and also called *Simple Update*.

In one dimensional system, the performance of Simple Update is pretty well, because 1-D systems obey the canonical form and have less influences of environment. However, in 2-D systems, due to the area law, we need consider the environment more restrictively when measuring the local observable. Moreover, the computational consumption is another serious problem, owing to the growth of a state's dimension which is proportional to dD^4 .

In order to solving that obstacles, optimizations of 2-D algorithms became an important part in condense matter physics. This chapter we focus on obtaining good enough projected entangled pair states from 2-D iTEBD and the strategies of improving measurement would be shown in following chapters.

3.2.2 Description and Pseudocode of iTEBD for 2-D system

Now that we stated to extend it to a two dimensional system. In chapter.??, we have known that a two dimensional many-body system is able to be represented by PEPS. Due to impossibly drawing an network of infinite sites, the structure of infinite PEPS (iPEPS) is decided by the geometry of the lattice and the unit cell we chosen. In usual, the size of unit cell depend on the n-site evolution operator. For instance, if the target is updating iPEPS of a square lattice with 2-site operator, the tensor diagram would be drawn as Fig.3.4.

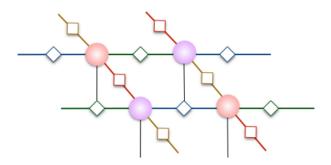


Figure 3.4: Four sites unit cell in iPEPS.

After setting the form of iPEPS, we start to deal the question of updating states. The most intuitive scheme is to apply the scheme of *Simple Update* which is shown in Fig.3.5.

The steps are similar to the iTEBD on 1-D systems. However, there are some differences. Firstly, the projected entangled pair states is a rank-5 tensor. Secondly, there are more entangled should be considered, due to increased neighbor sites.

More theoretical descriptions are written in [??][??]. Here, we explain the methods with TNs and some simple pseudocodes. The basic idea of 2-D iTEBD is to update the states from four directions by Simple Update. The example starting from updating the green bond is shown in Fig.3.5 and Fig.3.6,

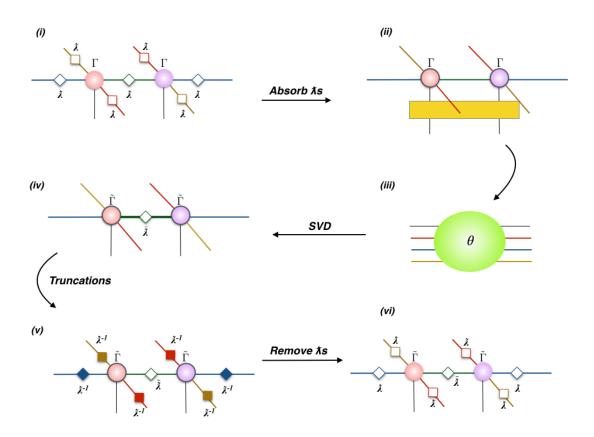


Figure 3.5: Absorb all λ to Γ . (ii) Contract an evolution operator $e^{-\delta H}$ for evolving the system. (iii) Decompose the tensor θ by SVD. (iv) Truncate and Update the states and λ on the green bond.(v) Remove λ for obtaining the states. (iv) Obtain a original form of iPEPS. Repeat all the step to update the other bonds until the ground state energy convergence

It's the same as one dimensional case, The first step is to absorb all λ around the sites. The tensor with gray bold means the state have absorbed all entanglements. Secondly, contract the gate, $e^{-\delta H}$, for getting the tensor θ . Thirdly, apply singular value decomposition to update states and the entanglement on the green bond. After decomposing θ , we found that the dimension of the green bond increase to dD^4 . Therefor, truncation plays

a significant roles for keeping the dimension in the algorithm. In the end of the updating processes, multiply pseudo inverse of all λ to the tensors for reducing to original form.

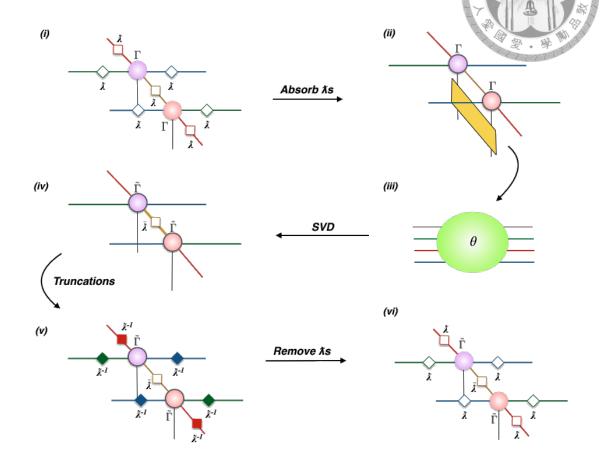


Figure 3.6: Update the yellow bond and the steps are similar to Fig.3.5

For easily to imagine how to updating the others directions. The updating steps of yellow bond are shown In Fig.3.6.

3.3 Ameliorate two-dimensional iTEBD

This method which make the algorithm more stable was published by *M. B. Hastings* [14]. Although *Simple Update* shown in section.3.2.2 can obtain pretty good states, it's not stable and efficient enough. The reason is that too many multiplications of pseudo inverse λ at the step Fig.3.5(v). In numerical methods, it's hard to deal the problem of dividing the value which is equal or approach to zero. On the other words, the more inverse operations, the more the probability of bring about divergence or destroying algorithms.

For reducing the risk of breaking algorithms. Firstly, declare the states Γ which include two entanglements among them. For instance, In Fig.3.7(i), the red tensor is considered as multiplication of A and the λ on the yellow. The purple one is multiplication of B and remained λ . Secondly, in Fig.3.7(ii), because the entanglements on red and blue bonds are included in tensor B, we just need absorb the yellow one and contract the evolution operation for obtaining tensor C. Thirdly, we contract the red and blue λ to the red and blue bonds which belong to the original tensor A in tensor C for getting θ . Fourthly, getting the B from decomposing and truncating θ . Owing to avoid multiplying inverse matrices, we get A by contracting tensor C and B and multiply an inverse λ of yellow bond for removing the entanglement and reducing to original form.

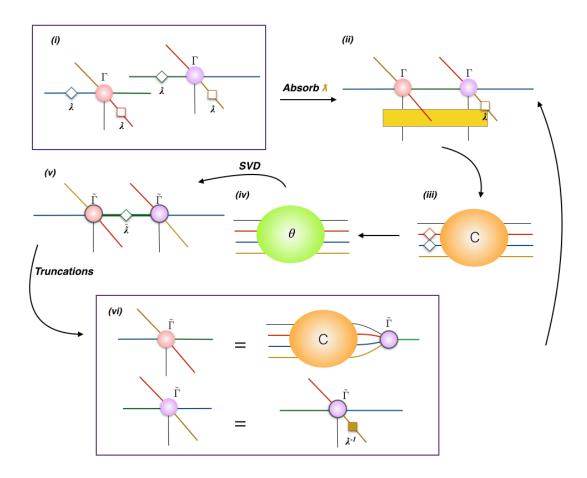


Figure 3.7: The tensor network diagrams for the ameliorated 2-D iTEBD.

Finally, repeat all the processes shown in Fig.3.7 to update different bonds until the convergence of the ground state energy.

3.4 Optimizations

3.4.1 Initialization



Intuitively, the initialization of states should not affect the result. However, it's a serious misunderstanding. Actually, stating from a awful initial sates may break the algorithms or hardly converge.

From the viewpoint of physics, translational invariance is one of essential properties in many-body system, So we can assume that the group state on two sites should be similar. For instance, if the TN diagram of the states is shown as Fig 3.8(i), Fig 3.8(ii) might be the better way to initialize the states.

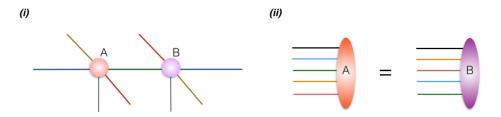


Figure 3.8: (i) The structure of PEPS, (ii) The initialization of states

3.4.2 Truncattion Error

3.4.3 QR decomposition

Though the strategy described in previous sections improve the stability, it's not efficient enough. The reason why is that the dimension of tensor θ , in Fig.3.5(iii) and Fig.3.7(iii), is proportional to d^2D^6 . In addition to that, the time complexity of singular value decomposition is proportional to $O(NM^2)$. In conclusion, those steps are expensive and the dimension of tensor θ must be reduced for accelerating.

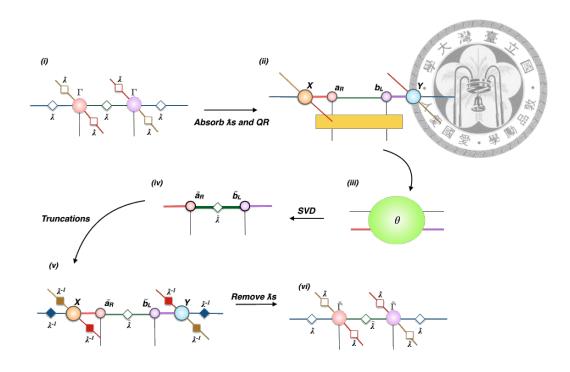


Figure 3.9: The tensor network diagrams for the ameliorated 2-D iTEBD.

To achieve the goal, the projected pair state must be decomposed by QR decomposition [13] [?]. The processes making *Simple Update* more efficient is illustrated in Fig 3.9. Most of the steps shown in Fig 3.9 are like in Fig 3.5. The only difference is that after absorbing all the λ , we decompose the state to an orthogonal matrix and an upper triangular matrix by QR. For instance, in Fig 3.9 (ii), the state A is decomposed to an orthogonal matrix X and upper triangular matrix A. Due to the columns of X are orthonormal, XX^{\dagger} is equal to X. In the other word, the tensor X can be ignored and we just need consider the tensor A by QR. Similarity, the state X can be decomposed to a lower triangular X and an orthogonal matrix X by LQ which is equivalence to operate QR decomposition after transpose the matrix. Next, (iii) we can obtain the tensor X whose dimension is X from X and an evolution operator.

The strategy to accelerate *Ameliorate Simple Update* is shown in the Fig 3.10 and its main idea is also to reduce the dimension of θ .

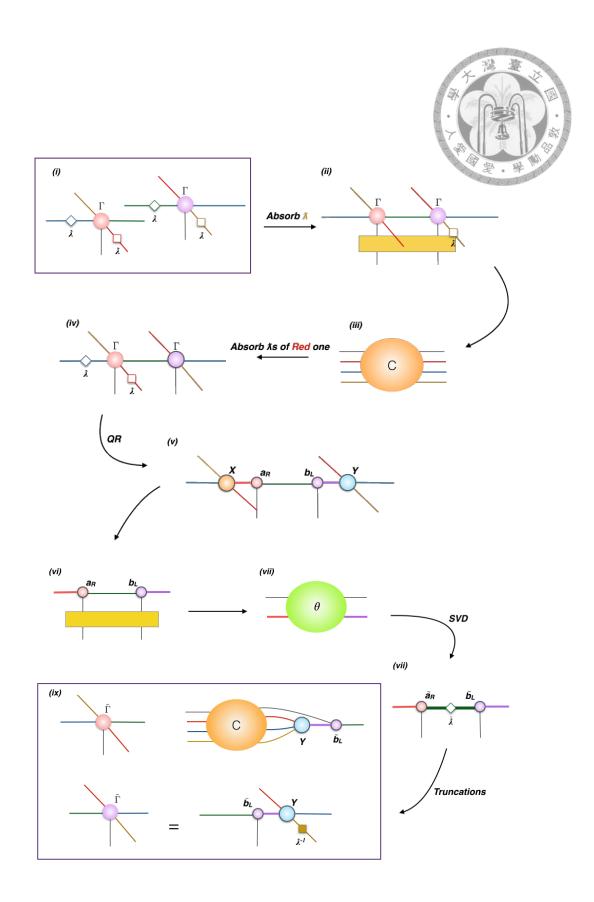


Figure 3.10: The tensor network diagrams for the ameliorated 2-D iTEBD.

3.5 Comparison

So far, we have benchmarked the improved iTEBD.



3.5.1 Different Initializations

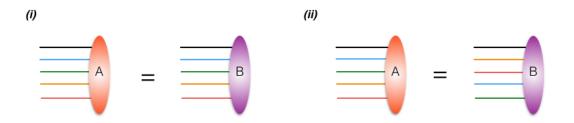


Figure 3.11: (i) Type 1, (ii) Type 2

See Fig 3.12, the both cases are

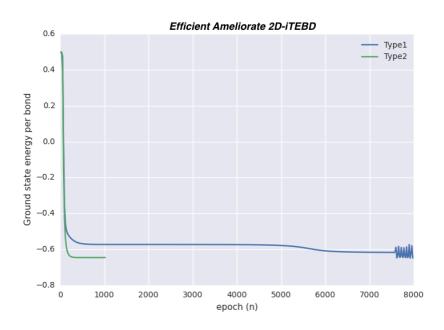


Figure 3.12: The Blue line represents updating tensors from the initial state shown in Fig 3.11 (i) and the green one represents updating from Fig 3.11 (ii)

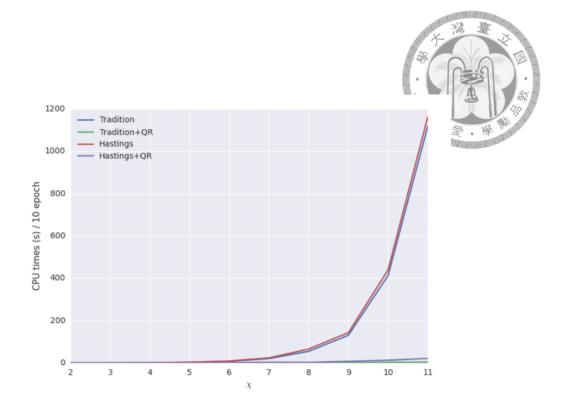


Figure 3.13

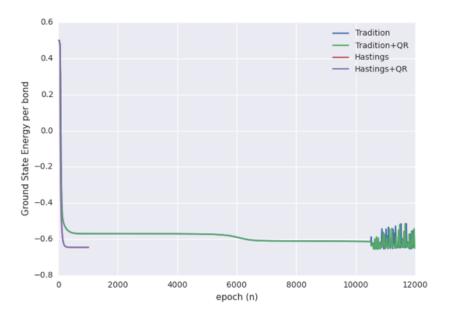


Figure 3.14

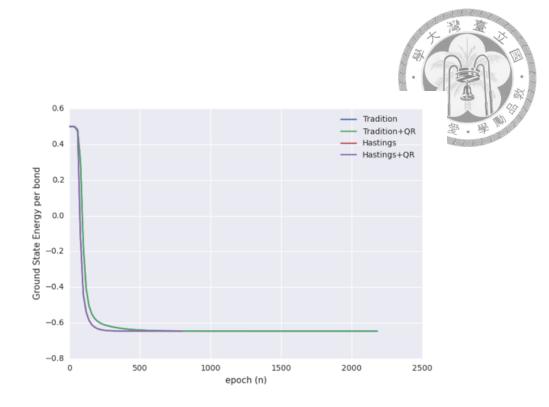


Figure 3.15

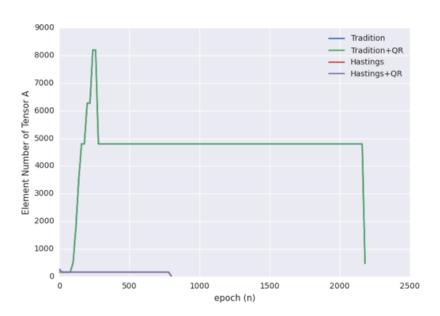


Figure 3.16



Chapter 4

Infinite Projected Entangled Simplex State

Introduce to PESS ansatz.

4.1 Simplex-solid State

The simplex-solid state of an SU(N) quantum anti-ferromagnet was derived by Arovas [] and the most significant conclusion is that any simplex states could be described by a natural generalization of the AKLT [] valence bond solid state which means that the bond signlets of the AKLT could be extended to N-site simplices.

The concept of simplex-solid states ansatz were introduce by Xie et al []. The tensor-network representation of simplex states is called projected entangled simplex state (PESS) which is also considered as an extension of PEPS [] obeying the area law of entanglement entropy and characterizing any states if the dimension of the virtual bonds are large enough. The difference from the PEPS is that the entanglement among the virtual particles is described by entangled simplex tensors which depend on the structure of PESS. For example, in Fig. 4.1(b) there are three virtual particles within the simplex state, so the entangled simplex tensor S_{mnl} is a rank-3 tensor.

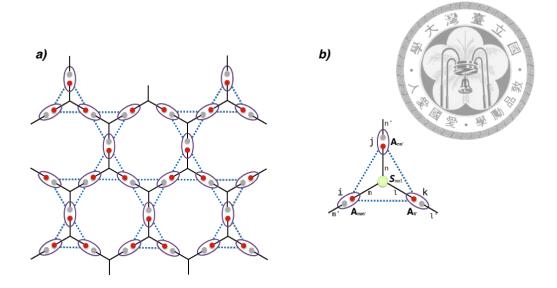


Figure 4.1: The red and blue tensor denotes on *odd* and *even* sites. The yellow one are time evolution operators $e^{-\tau H_{k,k+1}}$, $e^{-\tau H_{k+1,k}}$

4.2 Infinite Kagome Lattice

In this section, we will discuss about the formulation of the wave function of PESS and employ it as a variational ansatz. In the reference [], it has shown an example of the spin-2 simplex state on Kagome lattice for illustration. See Fig. 4.1(a) each physical S=2 states on the lattices could be treated as a symmetric superposition of two virtual S=1 spins. According to the theory of AKLT, each neighbor simplices (triangles) share a single site symmetrically which also means that the S=1 spins could be assigned to one of the simplices (vertex-sharing). Hence, there are three S=1 spins in each simplex states. From the features of the SU(2) group, the decomposition of the direct-product of three integer spins is written as,

$$n \otimes n \otimes n = [a_0 \times 0] \oplus \cdots \oplus [a_k \times k] \oplus \cdots \oplus [a_{3n} \times 3n]$$
 (4.1)

$$a_k = \begin{cases} 2k+1 & , k \le n \\ 3n+1-k & , k > n \end{cases}, k = 0, 1, \dots, 3n,$$
 (4.2)

where a_k is a constant and k express the kth irreducible representation. Now that we can write down the product of the spins in the simplex as,

$$\underline{1} \otimes \underline{1} \otimes \underline{1} = \underline{0} \oplus (3 \times \underline{1}) \oplus (2 \times \underline{2}) \oplus \underline{3},$$

and show that there is an unique spin-singlet state. The result encourage us to define a virtual singlet on simplex,

$$|\psi_{\alpha}\rangle = \frac{1}{\sqrt{6}} \sum_{\{s_i s_j s_k\}} S_{s_i s_j s_k}^{\alpha} |s_i, s_j, s_k\rangle, \tag{4.4}$$

where s_i, s_j, s_k are S=1 virtual spins located at site i, j and k containing in the simplex α and $S_{s_is_js_k}^{\alpha}$ is the Levi-Civita antisymmetric tensor ε_{ijk} [Fig. 4.1(b)]. For mapping the two virtual S=1 spins to the spin-2 subspace, we defined the projection operator P_i on each site,

$$P = \sum_{s_i, s_i'} \sum_{\sigma_i} A_{s_i, s_i'}^{\sigma_i} |\sigma_i\rangle \langle s_i, s_i'|$$

$$(4.5)$$

where $|\sigma_i\rangle$ is a basis of the S=2 spin at site i. $A^{\sigma_i}_{s_i,s_i'}$ is a projected matrix whose components are filled by the Clebsch-Gordan coefficients,

$$\begin{split} A_{11}^2 &= A_{33}^{-2} = 1, \\ A_{12}^1 &= A_{21}^1 = A_{23}^{-1} = A_{32}^{-1} = \frac{1}{\sqrt{2}}, \\ A_{13}^0 &= A_{31}^0 = \frac{1}{\sqrt{6}}, \\ A_{22}^0 &= \frac{2}{\sqrt{6}}, \end{split}$$

Now that we can write down the wave function of the simplex-solid state,

$$|\Phi\rangle = \bigoplus_{i} P_{i} \prod_{\alpha} |\psi_{\alpha}\rangle \tag{4.6}$$

$$= \operatorname{Tr}\left(\dots S_{s_i s_j s_k}^{\alpha} A_{s_i, s_i'}^{\sigma_i} A_{s_i, s_i'}^{\sigma_j} A_{s_k, s_k'}^{\sigma_k} \dots\right) \left|\dots \sigma_i \sigma_j \sigma_k \dots\right\rangle. \tag{4.7}$$

and apply the tensor-network representation to descrbe it, see in Fig. 4.1(b).

This structure could be extended to any higher integer spins. In conlusion, a physical

S=2n (even-integer) spin is considered as a symmetric superposition of two virtual S=n spins and S=2n-1 (odd-integer) is regarded as a symmetric superposition of a virtual S=n-1 spin and a virtual S=n spin [Fig. 4.1(a)]. More details are included in reference [] [].

4.3 Variational PESS ansatz

The PESS ansatz is similar to the imaginary time evolution discussed in chapter.3. However, unlike the PEPS algorithms, we apply a higher rank tensor S to describe the entanglement entropy among virtual particles in a simplex state. Due to the difference, we use *high-order singular value decomposition* (HOSVD) rather than SVD to decompose wave function.

4.3.1 High-order singular value decomposition

In the section, we will introduce to Nth-order singular value decomposition (HOSVD) which is proposed for decomposing rank-N tensors and show the pseudocode to illustrate the scheme of the implementation.

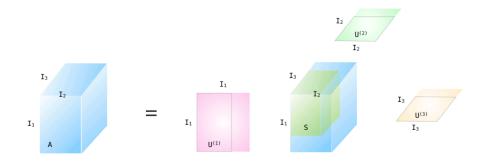


Figure 4.2: Picture of the HOSVD for a rank-3 tensor A. $U^{(1)},U^{(2)}$ and $U^{(3)}$ are unitary matrices and S (yellow cuboid) is a core tensor.

According to the theorem of HOSVD [], every *complex* $(I_1 \times I_2 \times \cdots \times I_N)$ -tensor A

could be decompose to a *core tensor* S and other n-mode unitary matrix $U^{(n)}$, where the maximum of n must be smaller than N,

$$A = S \times U^{(1)} \times U^{(2)} \times \dots \times U^{(n)}$$

As shown in Fig. 4.2, a rank-3 tensor A is decomposed to a core tensor S (yellow cuboid) and three unitary matrices obtaining from three different modes. The core tensor S is a complex ($I_1 \times I_2 \times \cdots \times I_N$)-tensor and have some significant properties,

1. Row-orthogonal: Two subtensors $S_{\alpha}^{(n)}$ and $S_{\beta}^{(n)}$ are orthogonal when $\alpha \neq \beta$.

$$\left\langle S_{\alpha}^{(n)}, S_{\beta}^{(n)} \right\rangle$$
 , if $\alpha = \beta$ (4.9)

In other words, no matter the shape of S, each rows are orthogonal.

2. Ordering: The Frobenius-norms $\left\|S_i^{(n)}\right\|$ is ordered from large to small.

$$||S_1^{(n)}|| \ge ||S_2^{(n)}|| \ge \dots \ge ||S_{I_n}^{(n)}|| \ge 0,$$
 (4.10)

for all possible values of n.

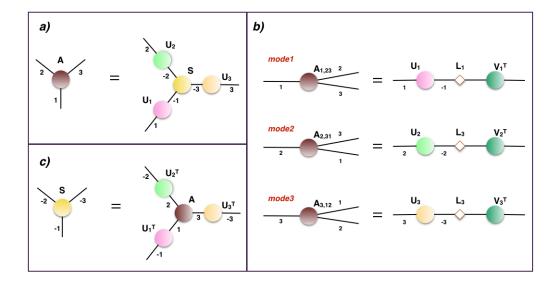


Figure 4.3: (a) Decompose A to a core tensor S and n-mode unitary matrix $U^{(n)}$, (b) Obtain tensors $U^{(n)}$ from decomposing various modes of tensor A by SVD, (c) Obtain the core tensor S from contracting all transpose unitary tensors $U^{(n)T}$

- 4.3.2 3-PESS
- 4.3.3 5-PESS



4.4 Infinite Square Lattice

The example on Kagome lattice.

- 4.4.1 4-PESS (Rank-3 local tensors)
- 4.4.2 4-PESS (Rank-5 local tensors)



Chapter 5

Corner Transfer Matrix

The cerner transfer matrix renormalization group (CTMRG) [] is an algorithm to numerically compute the effective environments which is an approximation of the environment of systems. For example, if the infinite PEPS is composed by a single tensor A^h_{uldr} repeatedly, where h express a physical basis of $\mathbb V$ with dimension d, and u, l, d, r are virtual bonds with dimension D, see Fig. 5.1(a). Then we can represent the scale norm $\langle \psi | \psi \rangle$ by a simple two dimensional tensor network ε which is characterized by reduced tensors a, shown in Fig. 5.1(b). The reduced tensor a is defined as eq.5.2,

$$a \equiv \sum_{h=1}^{d} A_h \otimes A_h^* \tag{5.1}$$

The environment $\varepsilon^{[\vec{r}]}$ of the site \vec{r} could be described by the reduced tensors in the gray rectangles in Fig.5.1(c) and the *effective environments* $G^{[\vec{r}]}$ shown in Fig. 5.1(d) is target of the CTMRG.

In the following subsections, we will show more details of implementation of CTM and compare some features between obtaining the states from iPEPE and PESS.

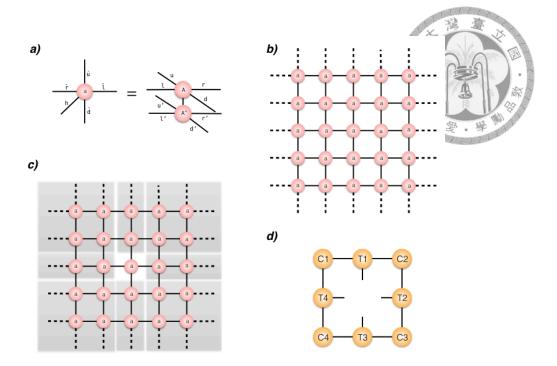


Figure 5.1: The red and blue tensor denotes on *odd* and *even* sites. The yellow one are time evolution operators $e^{-\tau H_{k,k+1}}$, $e^{-\tau H_{k+1,k}}$

5.1 Obtain States from PEPS

In chapter.3, we have discussed about obtaining the infinite PEPS state $|\psi\rangle$ of an infinite 2D square lattice by imaginary time evolution and known that the infinite PEPS could be characterized by two tensors A^h_{uldr} and B^h_{drul} repeatedly (Fig. 5.2(a)). The scaler norm of the iPEPS $\langle\psi|\psi\rangle$ is composed by reduced tensors $a_{\bar{u}\bar{l}d\bar{r}}$ and $b_{\bar{d}\bar{r}\bar{u}\bar{l}}$ (Fig. 5.2(b)), where

$$\bar{a} \equiv \sum_{h=1}^{d} A_h \otimes A_h^* \tag{5.2}$$

$$\bar{b} \equiv \sum_{h=1}^{d} B_h \otimes B_h^* \tag{5.3}$$

Then, we can consider the environment $\varepsilon^{[\vec{r_1},\vec{r_2},\vec{r_3},\vec{r_4}]}$ of a four-site structure (Fig. 5.2(c)), and try to approximate it with effective environment $G^{[\vec{r_1},\vec{r_2},\vec{r_3},\vec{r_4}]}$ (Fig. 5.2(d)), which consists of

$$C_1, T_{a1}, T_{b1}, C_2, T_{a2}, T_{b2}, C_3, T_{a3}, T_{b3}, C_4, T_{a4}, T_{b4},$$

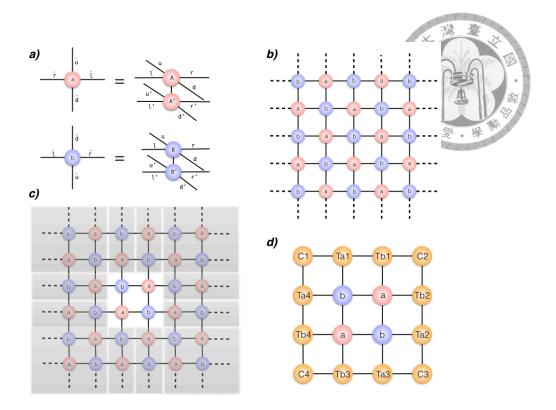


Figure 5.2: The red and blue tensor denotes on *odd* and *even* sites. The yellow one are time evolution operators $e^{-\tau H_{k,k+1}}$, $e^{-\tau H_{k+1,k}}$

For the approximation of environment, the directional variant of the CTMRG was developed. According to *directional coarse-graining moves*, the effective environment could be updated from four different moves, left, right, up and down and iterated until the environments converges.

For instance, the procedures to the left move, shown in the Fig. 5.3 which is derived by Roman and Vidal, is made up of four major steps,

- 1. Insertion: Insert two new columns which consist of $\{T_{a1}, b, a, T_{b3}\}$ and $\{T_{b1}, a, b, T_{a3}\}$ as in Fig. 5.3(b).
- 2. Absorption: In order to obtaining two new corner matrices \tilde{C}_1 and \tilde{C}_4 , and two new transfer matrices \tilde{T}_{b4} and \tilde{T}_{a4} , we contract tensors C_1 and T_{b1} , tensors C_3 and T_{a3} , tensors T_{a4} and b, and tensors T_{b4} and a. Then, contract tensors \tilde{C}_1 and \tilde{T}_{b4} , and tensors \tilde{C}_4 and \tilde{T}_{a4} , obtaining \tilde{Q}_1 and \tilde{Q}_4 which play significant rules for calculating isometries between \tilde{T}_{b4} and \tilde{T}_{a4} as in Fig. 5.3(c).

3. Renormalization: Truncate the vertical virtual bonds of \widetilde{C}_1 , \widetilde{T}_{b4} , \widetilde{T}_{a4} , and \widetilde{C}_4 by contracting the isometries Z and W, where

$$Z^{\dagger}Z = I$$

$$W^{\dagger}W = I \tag{5.5}$$

and the renormalization of the left CTM, yield as

$$C_1' = Z^{\dagger} \tilde{C}_1 \tag{5.6}$$

$$T'_{b4} = Z\tilde{T_{b4}}W^{\dagger} \tag{5.7}$$

$$T'_{a4} = W\tilde{T_{b4}}Z^{\dagger} \tag{5.8}$$

$$C_4' = Z\widetilde{C_4} \tag{5.9}$$

See the Fig. 5.3(d) and 5.3(f).

4. Truncation: To determinate the isometries Z and W in the *renormalization* steps is the most significant part. In this case, we use the eigenvalue decomposition of

$$\tilde{C}_1\tilde{C}_1^{\dagger} + \tilde{C}_4\tilde{C}_4^{\dagger} = \tilde{Z}D_z\tilde{Z}^{\dagger} \tag{5.10}$$

$$\tilde{Q}_1 \tilde{Q}_1^{\dagger} + \tilde{Q}_4 \tilde{Q}_4^{\dagger} = \tilde{W} D_w \tilde{W}^{\dagger} \tag{5.11}$$

shown in Fig. 5.3(e). It's not hard to find that the the dimension of bonds of D_z and D_w increase to χ^2 . For that reason, we have to truncate \tilde{Z} and \tilde{W} to isometries Z and W which are equivalent to keeping the columns of \tilde{Z} and \tilde{W} corresponding to χ largest eigenvalues of D_z and D_w .

Now, we need repeat the procedures in Fig. 5.3(b)-(d) again for absorbing the other inserted column in Fig. 5.3(d) and obtain a new effective environment $G'^{[\vec{r_1},\vec{r_2},\vec{r_3},\vec{r_4}]}$ for the four-site unit cell. By composing four variant moves of the CTM we build an epoch of CTMRG.

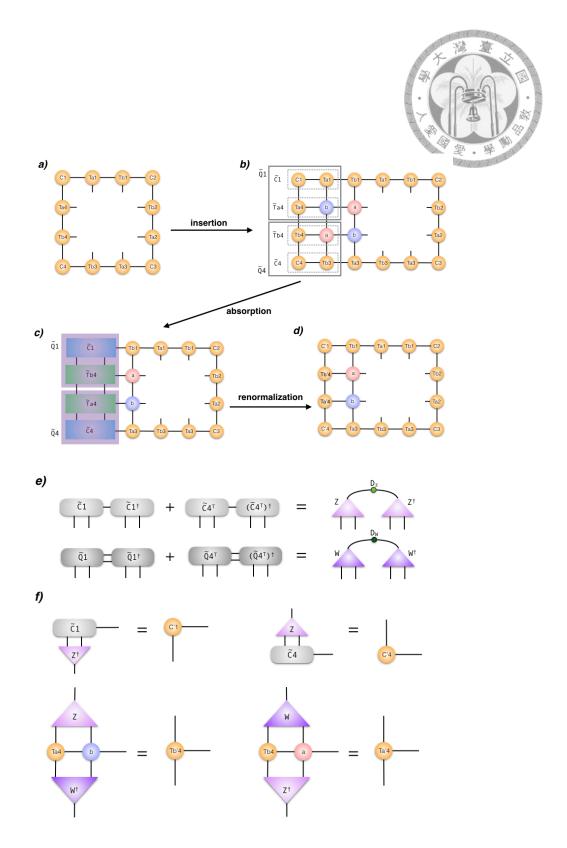


Figure 5.3: The red and blue tensor denotes on *odd* and *even* sites. The yellow one are time evolution operators $e^{-\tau H_{k,k+1}}$, $e^{-\tau H_{k+1,k}}$

5.2 Obtain States from PESS

CTM with PESS.





Chapter 6

Summary



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