### 1 Lecture 1

When investigating many-body quantum systems, we generally want statistical properties, even as our models are too complicated to solve exactly.

**Example** (Spin chain). A system of L spins has a Hilbert space  $\{|\uparrow\rangle, |\downarrow\rangle\}^{\otimes L}$  with dimension  $2^L$  making exact numerical calculations completely infeasible for  $L \gtrsim 30$ 



Instead we must change our mindset from finding information about all of the states, and instead focus only on the relevant sectors. For most purposes the states we are interested are near the *ground state* or *thermal states*. Hence we want a method to calculate states in these sectors, which is where the matrix product state (MPS) method comes into play.

Remark (MPS (PEPS/TNS) methods). Key features are:

- Works for 1D systems (For 2D, in developing phase yet!)
- Applicable for various system types like chains of bosons, fermions or spins (like Pott's chain etc)
- Works best for 'low' entanglement states (quantified later)
- There is primarily a possibility of finding the ground state and the low energy excitations
- It is possible to find time evolution for both *closed* and *open* systems
- The method also works with finite temperature states

# 1.1 Idea of DMRG/MPS (Variational method)

The matrix product state method is a variational method which relies heavily on singular value decompositions (SVD) and the Schmidt decomposition.

**Method** (Singular Value decomposition). Any rectangular matrix A of dimensions  $(m \times n)$  can be decomposed as

$$A = USV^{\dagger},$$

where the matrices  $U, S, V^{\dagger}$  are matrices with the below properties

 $U: A (m \times \min(m, n))$  matrix with  $U^{\dagger}U = I$ 

 $S: A (\min(m,n) \times \min(m,n))$  diagonal matrix with  $S_{\alpha\alpha} = \sqrt{\lambda_{\alpha}} \geq 0$ 

 $V^{\dagger}$ : A  $(\min(m, n) \times n)$  matrix with  $V^{\dagger}V = I$ 

The rank of S will turn out to be especially important and so in general we will denote r = rank(S).

**Method** (Schmidt decomposition). Using SVD we can decompose a general element of a product space  $\mathcal{H}_A \otimes \mathcal{H}_B$  from a double sum over tensor products of basis elements to a single sum over an orthonormal Schmidt basis as follows

$$|\psi\rangle = \sum_{ij} \psi_{ij} |i\rangle_A |j\rangle_B$$

$$= \sum_{ij\alpha} U_{i\alpha} \sqrt{\lambda_{\alpha}} V_{j\alpha}^* |i\rangle_A |j\rangle_B$$

$$= \sum_{\alpha=1}^r \sqrt{\lambda_{\alpha}} \left( \sum_i U_{i\alpha} |i\rangle_A \right) \left( \sum_j V_{j\alpha}^* |j\rangle_B \right)$$

$$= \sum_{\alpha=1}^r \sqrt{\lambda_{\alpha}} |\alpha\rangle_A |\alpha\rangle_B$$
(\*\*)

The Schmidt decomposition is exact and the coefficients  $\lambda_{\alpha}$  tell us very useful information about what happens if we trace out one of the two subsystems. It also provides the minimal number r of coefficients by which we may describe  $|\psi\rangle$  using a product basis.

#### Digression : $\rho$ after Schmidt decomp.

Although, we lose the familiarity of product basis set  $\{|i\rangle_A, |j\rangle_B\}$ , what we gain is the simplicity in the form of density matrix  $\rho$ .

$$\begin{split} \rho &= \left| \psi \right\rangle \left\langle \psi \right| \\ &= \sum_{ij} \sum_{kl} \psi_{kl}^* \psi_{ij} \left| i \right\rangle_A \left| j \right\rangle_B \left\langle k \right|_A \left\langle l \right|_B \end{split}$$
 after Schmidt decomp. 
$$= \sum_{\alpha,\beta=1}^r \sqrt{\lambda_\alpha \lambda_\beta} \left| \alpha \right\rangle_A \left| \alpha \right\rangle_B \left\langle \beta \right|_A \left\langle \beta \right|_B \end{split}$$

As we know, Tr  $\rho = 1$ , that implies the singular values of this particular SVD follow  $\sum_{\alpha}^{r} \lambda_{\alpha} = 1$ .

The point of matrix product states is now to find a way of implementing a variational technique for the states  $|\tilde{\psi}\rangle$  which takes the form of  $(\star)$  with only  $D \leq r$  terms. The point will be to vary the basis and coefficient to minimise  $||\tilde{\psi}\rangle - |\psi\rangle||$ . Without loss of generality we may order the coefficients  $\lambda_1 \geq \lambda_2 \geq \cdots \geq \lambda_r$ , which allows us to calculate the minimal distance

$$\left\| \left| \tilde{\psi} \right\rangle - \left| \psi \right\rangle \right\|^2 = 1 - \sum_{\alpha=1}^{D} \lambda_{\alpha}.$$

It is therefore self-evident that this is good approximation if  $\lambda_{\alpha}$  decay quickly.

# 1.2 Quantifying validity of approximation

By thinking of our total system as consisting of two subsystems, and tracing out one of them we can illuminate the meaning of the coefficients.

Suppose we are in a pure state of the total system so that  $\rho = |\psi\rangle\langle\psi|$ . The reduced density matrices for the subsystems may be found by tracing out the other part of the Hilbert space

$$\rho_A = \operatorname{Tr}_B \rho,$$
  
$$\rho_B = \operatorname{Tr}_A \rho.$$

The coefficients  $\lambda_{\alpha}$  are therefore the eigenvalues of the reduced density matrices. A measure of how mixed the state is the von-Neumann entropy

$$S = -\operatorname{Tr} \rho_A \ln \rho_A = -\sum_{\alpha} \lambda_{\alpha} \ln \lambda_{\alpha},$$

which is small when  $\lambda_{\alpha}$  decay fast, and maximal when they are all equal. Hence the approximation is good if S is small.

**Remark.** From this we may conclude that MPS methods approximates the true state well for low entropy states. An equivalent way of viewing these are as states with a low amount of entanglement between the subsystems<sup>1</sup>

**Remark.** Some general theorems about the growth of entropy as system size increases are known

- Ground state of 1D gapped system with short range interaction  $S \to \text{const}$  as  $L \to \infty$ . Hence we only require our approximate state to have rank  $D \sim 2^{\text{const}}$  indepedent of system size.<sup>2</sup>
- Ground state of 1D critical system  $S \to R \ln L + \text{const}$  need  $D \sim L^R$ , which is polynomial in L, and therefore much more tractable than the exponential growth we see in the dimension of the full Hilbert space.

### 1.3 What are MPS?

So far we have no idea how we are going to implement our minimisation technique, or where the matrix product states come into play. Any  $|\psi\rangle \in \mathcal{H}^{\otimes L}$  state can be decomposed into a so-called MPS by essentially unfolding the coefficients of into a product of matrices. Let

$$|\psi\rangle = \sum_{\sigma_1 \cdots \sigma_L} C_{\sigma_1, \dots, \sigma_L} |\sigma_A\rangle \otimes \dots \otimes |\sigma_L\rangle$$

<sup>&</sup>lt;sup>1</sup>Ask lecturer to clarify that it is the entanglement between two subsystems which is small <sup>2</sup>Shouldn't the ground state be a pure state? Ask follow up, if means ground state of combined system A + B.

be a representation of any given state. Now define a matrix  $\psi$  of dim  $d \times d^{L-1}$  where  $d = \dim \mathcal{H}$  by the relation

$$\psi_{\sigma_1,(\sigma_2,\ldots,\sigma_L)} = C_{\sigma_1\sigma_2\ldots\sigma_L}$$

Clearly we can do SVDs iteratively on  $\psi$ 

$$\begin{split} \psi_{\sigma_{1},(\sigma_{2},\ldots\sigma_{L})} &= \sum_{a_{1}=1}^{r_{1}} U_{\sigma_{1},a_{1}} S_{a_{1},a_{1}}(V^{\dagger})_{a_{1},(\sigma_{2},\ldots,\sigma_{L})} \\ &= \sum_{a_{1}=1}^{r_{1}} A_{a_{1}}^{\sigma_{1}} C_{a_{1},\sigma_{2},\ldots\sigma_{L}} \\ &= \sum_{a_{1}=1}^{r_{1}} \sum_{a_{2}}^{r_{2}} A_{a_{1}}^{\sigma_{1}} U_{(a_{1},\sigma_{2}),a_{2}} S_{a_{2}a_{2}}(V^{\dagger})_{a_{2},(\sigma_{3}\ldots\sigma_{L})} \\ &= \sum_{a_{1},a_{2}} A_{a_{1}}^{\sigma_{1}} A_{a_{1}a_{2}}^{\sigma_{2}} \psi_{(a_{2},\sigma_{3})(\sigma_{4}\cdots)} \\ &= \sum_{a_{1}\ldots a_{L-1}} A_{a_{1}}^{\sigma_{1}} A_{a_{1}a_{2}}^{\sigma_{2}} \cdots A_{a_{L-2}a_{L-1}}^{\sigma_{L-1}} A_{a_{L-1}}^{\sigma_{L}} \end{split}$$

Where we've decomposed the tensor coefficient into a product of matrices of dimension  $(1 \times d)(d \times d^2) \dots (d^{L/2-1} \times d^{L/2})(d^{L/2} \times d^{L/2-1}) \dots (d \times 1)$ . It can be nice to think about what we are doing in diagrams, by representing a tensor as a square with lines corresponding to each index. Our matrix product state then is constructed by pulling apart each of the external indices as

