MULTI-CONFIGURATION TIME DEPENDENT HARTREE THEORY

A TENSOR NETWORK PERSPECTIVE

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Outline

Discrete Variable Representation

Multi-Configuration Time Dependent Hartree

MCTDH vs. DMRG

DISCRETE VARIABLE REPRESENTATION

DISCRETE VARIABLE REPRESENTATION

PROBLEM DESCRIPTION

Non-Relativistic Multi-Dimentional Problem

· TDSE:

$$\mathrm{i}\hbar\frac{\partial}{\partial t}\left|\Psi(\vec{q},t)\right\rangle=H|\Psi(\vec{q},t)\rangle$$

· TISE:

$$H|\Psi(\vec{q},t)\rangle = E|\Psi(\vec{q},t)\rangle$$

where

$$H = T + V$$

and $\vec{q} = (q_1, q_2, \dots, q_d)$.

DISCRETE VARIABLE REPRESENTATION

STANDARD PROCEDURE

Discretization: Spectral Method

Choose a set of basis functions: $\{|\Phi_I(\vec{q})\rangle\}_{I=1}^N$ and require

- 1. Orthonormal: $\langle \Phi_I | \Phi_J \rangle = \delta_{IJ}$
- 2. Complete: $\sum_{I=1}^{n} |\Phi_I\rangle\langle\Phi_I| = P \to 1, \quad n \to +\infty$

Therefore,

$$|\Psi\rangle pprox P|\Psi\rangle = \sum_{I}^{n} c_{I} |\Phi_{I}\rangle$$
 where $c_{I} = \langle \Phi_{I} | \Psi \rangle$

and

$$A \approx PAP = \sum_{IJ}^{n} |\Phi_{I}\rangle A_{IJ} \langle \Phi_{J}|$$
 where $A_{IJ} = \langle \Phi_{I}|A|\Phi_{J}\rangle$

http://www.pci.uni-heidelberg.de/tc/usr/mctdh/lit/NumericalMethods.pdf

1-D Problem

Standard procedure:

- 1. Choose a set of 1-D basis $\{|\varphi_i(x)\rangle\}_{i=1}^n$
- 2. Integrate

$$T_{ij} = \langle \varphi_i(x) | T | \varphi_j(x) \rangle = -\frac{\hbar}{2\mu} \int \varphi^*(x) \frac{\mathrm{d}}{\mathrm{d}x} \varphi(x) \, \mathrm{d}x,$$

$$V_{ij} = \langle \varphi_i(x) | V | \varphi_j(x) \rangle = \int \varphi^*(x) \, V(x) \varphi(x) \, \mathrm{d}x, \text{ and }$$

$$H_{ij} = T_{ij} + V_{ij}$$

- 3. Solve TDSE/TISE in matrix form
 - · TDSE:

$$\mathrm{i}\hbar\dot{c}_i=\sum_j H_{ij}c_j$$
 or $\mathrm{i}\hbar\dot{\mathbf{c}}=\mathbf{H}\mathbf{c}$

· TISE:

$$\sum_{j} H_{ij} c_j = E c_i \quad \text{or} \quad \mathbf{Hc} = E \mathbf{c}$$

d-D Problem: $\vec{q} = (q_1, \dots, q_d)$

Standard procedure:

- 1. Choose a set of $d ext{-D}$ basis $\{|\Phi_I(\vec{q})\rangle\}_{I=1}^N$
 - \cdot Use the direct product of d sets of 1-D basis

$$\bigotimes_{\kappa=1}^{d} \left\{ \left| \varphi_i^{(\kappa)}(q_\kappa) \right\rangle \right\}_{i=1}^{n_\kappa}$$

as the set of d-D basis, i. e.,

$$|\Phi_I(q_1,\ldots,q_d)\rangle = \prod_{\kappa=1}^d \left|\varphi_{i_\kappa}^{(\kappa)}(q_\kappa)\right\rangle$$

where $I = (i_1, \ldots, i_d)$.

$$d$$
-D Problem: $\vec{q}=(q_1,\ldots,q_d)$

Standard procedure:

- 1. Choose a set of d-D basis $\{|\Phi_I(\vec{q})\rangle\}_{I=1}^N$, where $|\Phi_I(\vec{q})\rangle = \prod_{\kappa=1}^d \left|\varphi_{i_\kappa}^{(\kappa)}(q_\kappa)\right\rangle$
- 2. Integrate $H_{IJ} = T_{IJ} + V_{IJ}$
- 3. Solve TDSE/TISE in matrix form
 - · TDSE: $\mathrm{i}\hbar\dot{\mathbf{c}}=\mathbf{H}\mathbf{c}$
 - TISE: $\mathbf{Hc} = E\mathbf{c}$

DISCRETE VARIABLE REPRESENTATION

SELECTION OF BASIS

Motivation of DVR

- o How to do the integral in the standard procedure?
- Kinetic energy
 In Cartesian coordinate

$$T = -\frac{\hbar}{2} \sum_{\kappa=1}^{d} \frac{1}{\mu_{\kappa}} \frac{\partial^{2}}{\partial q_{\kappa}^{2}}$$

Therefore,

$$\mathbf{T} = \bigotimes_{\kappa=1}^{d} \mathbf{T}^{(\kappa)} \quad \text{where} \quad T_{ij}^{(\kappa)} = -\frac{\hbar}{2\mu_{\kappa}} \left\langle \varphi_{i}^{(\kappa)} \left| \frac{\partial^{2}}{\partial q_{\kappa}^{2}} \right| \varphi_{j}^{(\kappa)} \right\rangle$$

which can be integrated analytically once the set of basis is determined.

Motivation of DVR

- o How to do the integration in the standard procedure?
- Kinetic energy
- Potential energy
 Differs in different problems.
- Can we integrate some physical quantity analytically once the set of basis is determined, and transform to the potential matrix without integration?

DVR

Fact

If
$$\langle \Phi_I | \vec{x} | \Phi_J \rangle = \vec{x}_I \delta_{IJ}$$
 and $A = A(\vec{x})$, then $\langle \Phi_I | A | \Phi_J \rangle = A(\vec{x}_I) \delta_{IJ}$.

For a given set of basis functions $\{|\Xi_I(\vec{q})\rangle\}_{I=1}^N$, let $\vec{q}_{IJ} = \langle \Xi_I | \vec{q} | \Xi_J \rangle$, then there is a set of basis functions $\{|\Phi_I(\vec{q})\rangle\}_{I=1}^N$ s.t.

- 1. $|\Phi_J\rangle = \sum_{I=1}^N |\Xi_I\rangle \; U_{IJ}$, where **U** is unitary, and
- 2. $\vec{\mathbf{Q}} = \mathbf{U}^{\dagger} \vec{\mathbf{q}} \mathbf{U}$, where $\vec{Q}_{IJ} := \langle \Phi_I | \vec{q} | \Phi_J \rangle = \vec{q}_I \delta_{IJ}$.

DVR

- $\{|\Xi_I(\vec{q})\rangle\}_{I=1}^N$: finite basis set representation
- $\{|\Phi_I(\vec{q})\rangle\}_{I=1}^N$: discrete variable representation

In DVR basis, $V_{IJ} = V(\vec{q}_I)\delta_{IJ}$.

- \circ How to calculate $T|\Psi\rangle$?
- Integrate in FBR and transform in DVR $(\mathcal{O}(\mathit{N}^{2}))$
- Use FFT and its inverse to calculate the differential $(\mathcal{O}(N\log N))$

Example: (1-D) Sine-DVR Basis

Choose the n-th lowest eigenstates of 1-D infinite square well as FBR:

$$\xi_n(x) = \begin{cases} \sqrt{\frac{2}{L}} \sin(n\pi x/L), & \text{if } 0 \leqslant x \leqslant L \\ 0, & \text{else} \end{cases}$$

and in FBR, we have

$$T_{ij}^{(\text{FBR})} = -\frac{\hbar^2}{2\mu} \left(\frac{i\pi}{L}\right)^2 \delta_{ij}$$

Example: (1-D) Sine-DVR Basis

Let $q = \cos(\pi x/L)$, then \mathbf{q} could be diagonalized analytically:

$$U_{ij}=\sqrt{rac{2}{n+1}}\sinigg(rac{ij\pi}{n+1}igg)$$
 $Q_{ij}=q_i\delta_{ij}$ where $q_i=rac{iL}{n+1}$

and $\mathbf{Q} = \mathbf{U}^{\dagger} \mathbf{q} \mathbf{U}$.

Example: (1-D) Sine-DVR Basis

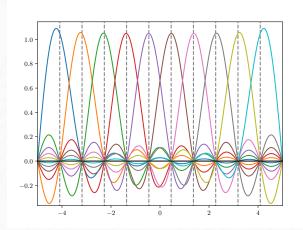


Figure 1: Sine-DVR basis functions (L=10 a. u., n=10)

Example: 1-D Double Well

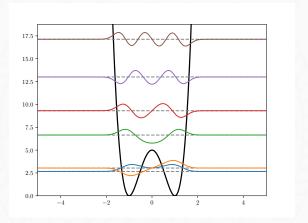


Figure 2: Potential curve and eigenstates

$$V = 5(x^2 - 1)^2$$
, $L = 10$, $n = 400$ (in a.u.)

Example: 1-D Double Well

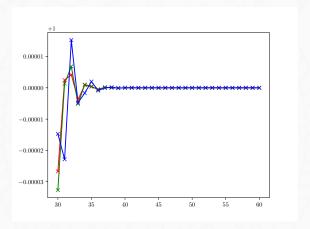


Figure 3: Relative energy to n

Red: ground state; green: 1st excited state; blue: 2nd excited state. Relative to when $L=10,\,n=400$

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Example: 1-D Double Well

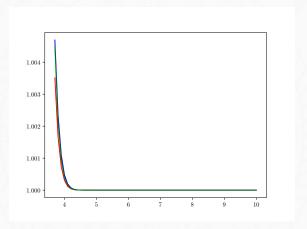


Figure 4: Relative energy to L

Red: ground state; green: 1st excited state; blue: 2nd excited state. Relative to when $L=10,\,n=400$

Example: 2-D harmonic oscillator

$$V = \frac{1}{2}x^2 + \frac{1}{2}y^2 + \frac{1}{4}xy, \quad n = 40, \quad L = 10 \quad \text{(in a. u.)}$$
 Start from the ground state when $V = \frac{1}{2}x^2 + \frac{1}{2}y^2$.

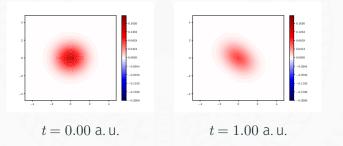
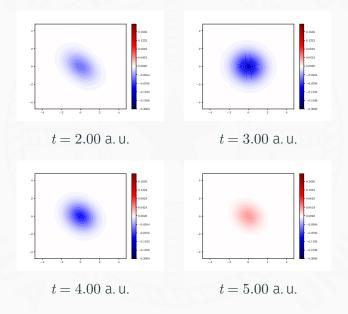


Figure 5: Real parts of wavefunctions of a 2-D harmonic oscillator at t (using RK45 as ODE solver, $\Delta t = 0.01$ a.u.)

Example: 2-D harmonic oscillator (Cont'd)



Example: 2-D harmonic oscillator

Autocorrelation function: $a(t)=\langle \Psi(0)|\Psi(t)\rangle$, and when initial state $|\Psi(0)\rangle$ is real, $a(t)=\langle \Psi^*(t/2)|\Psi(t/2)\rangle$

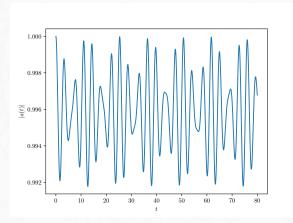


Figure 6: |a(t)| - t (using RK45 as ODE solver, $\Delta t = 0.001$ a. u.)

MULTI-CONFIGURATION TIME DEPENDENT HARTREE

Language: Tensor Network Notation

Tensors



Contraction

MULTI-CONFIGURATION TIME DEPENDENT HARTREE

MOTIVATION

Express the Standard Procedure in TNN

Wavefunction

$$\langle \Phi_I | \Psi \rangle \quad \stackrel{i_1}{=} \quad \stackrel{i_2}{\swarrow} \quad \stackrel{i_d}{\longrightarrow} \quad \stackrel{$$

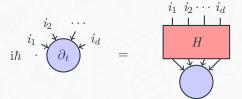
where $i_{\kappa} \in \{1, \dots, n_{\kappa}\}, \kappa = 1, \dots, d$. Space for saving a wavefunction: $\prod_{\kappa=1}^{d} n_{\kappa}$ floats.

Express the Standard Procedure in TNN

· Normalization condition

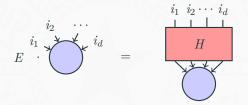
$$1 = \langle \Psi | \Psi \rangle \approx \langle \Psi | P | \Psi \rangle =$$

· TDSE



Express the Standard Procedure in TNN

TISE



where
$$i_{\kappa} \in \{1, \ldots, n_{\kappa}\}, \kappa = 1, \ldots, d$$
.

Problem: $N = \prod_{\kappa=1}^d n_{\kappa}$ exponentially increase as d grows.

Motivation of MCTDH

Motivation: construct the wavefunction without form the whole space.

One simple way: Hartree method.

http://www.pci.uni-heidelberg.de/tc/usr/mctdh/lit/intro_MCTDH.pdf

Hartree method in TNN

$$\langle \Phi_I | \Psi \rangle = \qquad \underbrace{i_1 \downarrow i_2 \downarrow \cdots \downarrow i_d \downarrow}_{i_1 \downarrow i_2 \downarrow \cdots \downarrow i_d \downarrow}$$

where

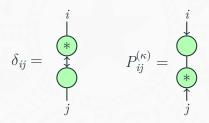
$$\left\langle \Phi_J^{(1)} \middle| \Psi \right\rangle = \int_0^{j_1} \int_0^{j_2} \cdots \int_0^{j_d} dt$$

and $j_{\kappa} \in \{1\}, \kappa = 1, \ldots, d$.

Space for saving a wavefunction: $\sum_{\kappa=1}^{d} n_{\kappa}$ floats.

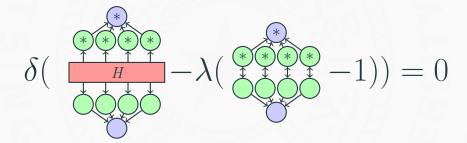
Hartree method in TNN

Extra normalization condition at the κ -th set of single particle functions (SPFs):



Variational Principle

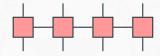
Solve TISE



Only special "structure" of ${\bf H}$ can simply the equation.

Structure of H

Matrix product operators (MPOs)



Summation of products of operators



MULTI-CONFIGURATION TIME DEPENDENT HARTREE

DATA STRUCTURE

Structure of Wavefunction

$$\langle \Phi_I | \Psi \rangle \quad = \quad \underbrace{i_1 \downarrow i_2 \downarrow \cdots \downarrow i_d \downarrow}_{l_2 \downarrow \cdots \downarrow l_d \downarrow}$$

where

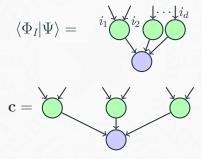
$$\left\langle \Phi_{J}^{(1)} \middle| \Psi \right\rangle = \stackrel{j_1 \quad j_2 \quad \dots \quad j_d}{}$$

Idea: $j_{\kappa} \in \{1, \ldots, n_{\kappa}^{(1)}\}, \kappa = 1, \ldots, d$, and $n_{\kappa}^{(1)} < n_{\kappa}$.

Space for saving a wavefunction: $\prod_{\kappa=1}^d n_{\kappa}^{(1)} + \sum_{\kappa=1}^d n_{\kappa}^{(1)} n_{\kappa}$ floats.

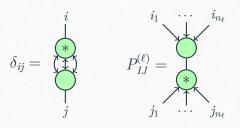
Mode Combination

Examples:



Normalization

General normalization condition at the ℓ -th node:



Multi-Layer

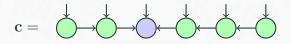
More nodes, but smaller ranks.

Example 1:

$$\mathbf{c} = \begin{array}{c} \\ \\ \\ \end{array}$$

Multi-Layer

Example 2: matrix product states (MPSs) in DMRG



For a complete binary tree, the space for saving a wavefunction is $\mathcal{O}(dn^3)$ floats, if $n_\ell = \mathcal{O}(n)$ for all ℓ .

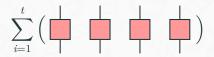
Generally, if rank $\leq p$ for all nodes, the space for saving a wavefunction is of $\mathcal{O}(dn^p)$.

MULTI-CONFIGURATION TIME DEPENDENT HARTREE

ALGORITHM

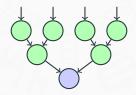
Solve TDSE

Use the same type of **H** in Hartree method:



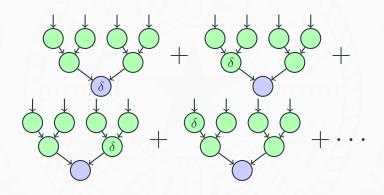
Dirac-Frenkel variational principle: $\langle \delta \Psi | \mathrm{i}\hbar \frac{\partial}{\partial t} - H | \Psi \rangle = 0.$

Without loss of generality, let $\mathbf{c}=$



Solve TDSE

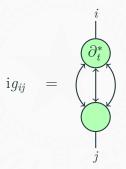
$$\delta \mathbf{c} =$$



Similarly, $\dot{\mathbf{c}}$.

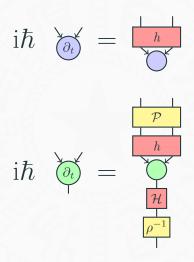
Normalization

In order to hold all general normalization conditions during the propergation, one must have



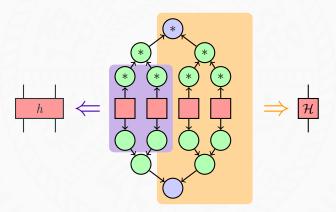
where \mathbf{g} is Hermitian. For simplicity, choose $\mathbf{g} = 0$.

Equations of Motion



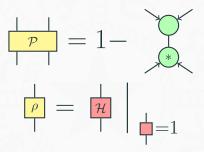
Equations of Motion (Cont'd)

where for all nodes, w.l.o.g., choose one term in ${\bf H}$ and one node as an example:



Equations of Motion (Cont'd)

and



Time complexity: if rank $\leq p$ for all nodes, then a step of multiplication is of $\mathcal{O}(tpdn^{p+1})$;

Space complexity: $\mathcal{O}(tpdn^2 + dn^p)$.

Example: 2-D harmonic oscillator

$$V = \frac{1}{2}x^2 + \frac{1}{2}y^2 + \frac{1}{4}xy$$
 (all in a. u.).

Start from the ground state when $V = \frac{1}{2}x^2 + \frac{1}{2}y^2$. Use Sine-DVR as the (primitive) basis set and n = 40, L = 10.

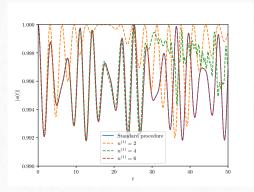


Figure 7: |a(t)| - t (using RK45 as ODE solver, $\Delta t = 0.001$ a. u.)

MCTDH vs. DMRG

MCTDH vs. DMRG

DATA STRUCTURE: A TN PERSPECTIVE

Structure of Wavefunction

· MCTDH

$$\mathbf{c} =$$

· DMRG

MCTDH vs. DMRG

ALGORITHM: DIFFERNET CHOICES

Algorithms

	TDSE	TISE
DMRG	Based on propergators	DMRG1, DMRG2
MCTDH	Based on DFVP	Self-consistent

o Other combinations?

Renormalization

The essence of sweeping over the tensor train.



Realized by (compressed) singular value decomposition (SVD).

- Use it in other TN-type wavefunctions.
- Equations often get simpler at the root node.

Nakatani, N. & Chan, G. K.-L. J. Chem. Phys. 138, 134113 (2013). (arXiv:1302.2298)

Another Problem

 How to determine the TN structure of a wavefunction of a specific Hamiltonian a priori?

Conclusions

- The structure of wavefunctions in MCTDH and DMRG can be unified in tensor network theory;
- The algorithms used in traditional MCTDH and DMRG are interchangeable in principle;
- The proper structure of wavefunctions in a specific problem needs further studying.

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Thank you for your listening!