# 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: $\vec{q} = (q_1, \ldots, q_d)$

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

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

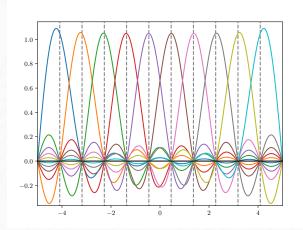
## Example: (1-D) Sine-DVR Basis

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

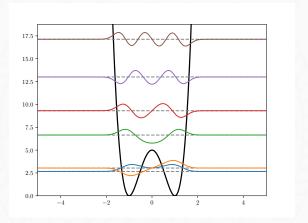
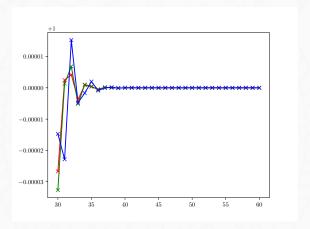


Figure 2: Potential curve and eigenstates

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

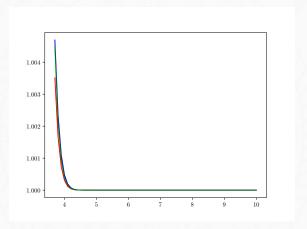


Figure 4: Relative energy to L

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  $V = \frac{1}{2}x^2 + \frac{1}{2}y^2$ .

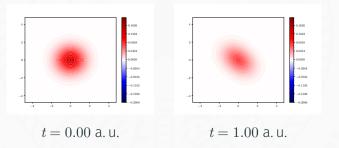
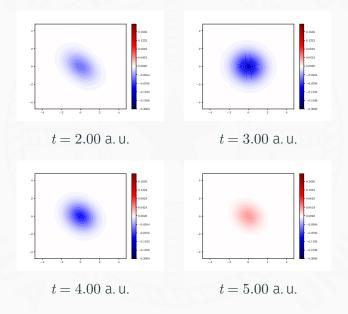


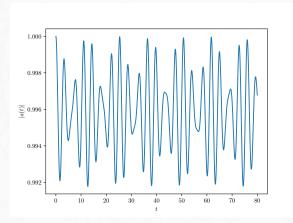
Figure 5: Wavefunctions of a 2-D harmonic oscillator at t (using RK45 as ODE solver,  $\Delta t = 0.01$  a. u.)

# 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

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

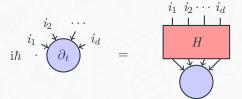
where  $i_{\kappa} \in \{1, \dots, n_{\kappa}\}, \kappa = 1, \dots, d$ . Space for saving a wavefunction:  $\prod_{\kappa=1}^{d} n_{\kappa}$  floats.

### Express the Standard Procedure in TNN

· Normalization condition

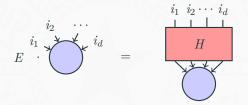
$$1 = \langle \Psi | \Psi \rangle \approx \langle \Psi | P | \Psi \rangle =$$

· 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/introMCTDH.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

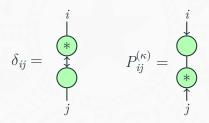
$$\left\langle \Phi_J^{(1)} \middle| \Psi \right\rangle = \int_0^{j_1} \int_0^{j_2} \cdots \int_0^{j_d} dt$$

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

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

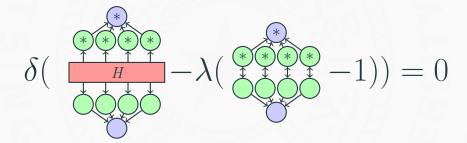
### Hartree method in TNN

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



## Variational Principle

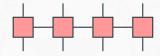
Solve TISE



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

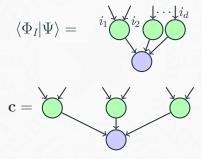
$$\left\langle \Phi_{J}^{(1)} \middle| \Psi \right\rangle = \stackrel{j_1 \quad j_2 \quad \dots \quad j_d}{}$$

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.

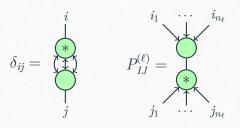
### **Mode Combination**

#### Examples:



### Normalization

General normalization condition at the  $\ell$ -th node:



# Multi-Layer

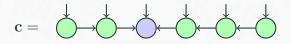
More nodes, but smaller ranks.

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  $\ell$ .

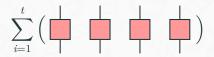
Generally, if rank  $\leq p$  for all nodes, the space for saving a wavefunction is of  $\mathcal{O}(dn^p)$ .

# MULTI-CONFIGURATION TIME DEPENDENT HARTREE

**ALGORITHM** 

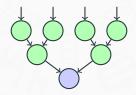
#### Solve TDSE

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



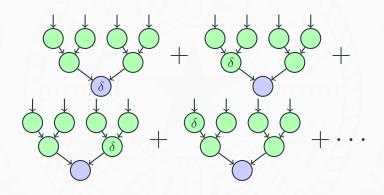
Dirac-Frenkel variational principle:  $\langle \delta \Psi | \mathrm{i}\hbar \frac{\partial}{\partial t} - H | \Psi \rangle = 0.$ 

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



# Solve TDSE

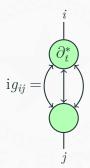
$$\delta \mathbf{c} =$$



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

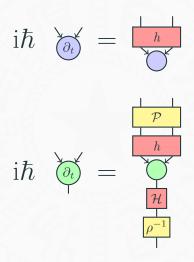
#### Normalization

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



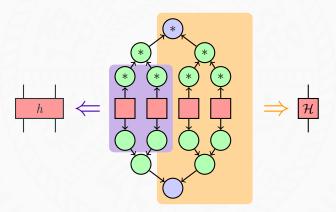
where  $\mathbf{g}$  is Hermitian. For simplicity, choose  $\mathbf{g} = 0$ .

# **Equations of Motion**



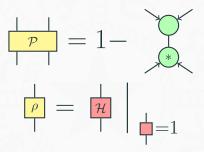
# Equations of Motion (Cont'd)

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



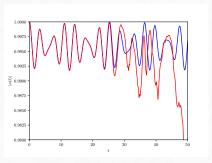
Time complexity: if rank  $\leq p$  for all nodes, then a step of multiplication is of  $\mathcal{O}(tpdn^{p+1})$ ;

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., renormalize all nodes during the propergation.) **Red:** MCTDH; blue: Standard procedure.

# 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.

## Acknowledgments

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Thank you for your listening!