

# Selected Configuration Interactions + Perturbation: Theory and Parallel Implementations

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# Goal of the Talk

- Introduction on SCI+PT methods for ground state calculations
- Discussion on SCI+PT parallel implementations
- Not included in this talk:
  - SCI+PT methods for excited states
  - Comparison with other CI methods (e.g. FCIQMC)
  - Where to eat dinner tonight

# Outline

- 1 Background
- 2 SCI+PT Methods
  - Selected CI: How to select?
  - Perturbation: How to perturb?
- 3 Parallel Implementations
  - Load Balancing
  - Reduce MPI Communications

## 1 Background

## 2 SCI+PT Methods

- Selected CI: How to select?
- Perturbation: How to perturb?

## 3 Parallel Implementations

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# The Many-Body Time-Independent Schrödinger Equation

- Ground-state wavefunction  $|\Phi_0\rangle$  and energy  $E_0$  of a chemical system given by

$$\hat{H}|\Phi_0\rangle = E_0|\Phi_0\rangle,$$

where  $\hat{H} = -\frac{1}{2} \sum_{i=1}^{n_{\text{elec}}} \nabla_i^2 + V(x_1, x_2, \dots, x_{n_{\text{elec}}})$ .

- Application: Molecular property calculations
- Computational Methods:
  - Density Functional Theory (DFT)
  - Many-Body Perturbation Theory (MBPT)
  - Coupled Cluster (CC)
  - **Full Configuration Interaction (FCI)**: for strongly-correlated system

# Configuration Interaction Method

- Based on one-electron spin-orbitals  $\{\chi_p\}_{p=1}^{n_{\text{orb}}}$  from Hartree-Fock procedure
- Wavefunction approximated as linear combination of anti-symmetrized tensor products (Slater determinants)

$$|\Phi_0\rangle = \sum_{i=1}^{N_{\text{FCI}}} c_i |D_i\rangle = \sum_{i=1}^{N_{\text{FCI}}} c_i |\chi_{p_i} \chi_{p_j} \cdots \chi_{p_k}\rangle$$

- FCI variational space dimension:  $N_{\text{FCI}} = \binom{n_{\text{orb}}}{n_{\text{elec}}}$
- Schrödinger equation transformed to eigenvalue problem

$$H\mathbf{c} = E_0\mathbf{c}, \quad H \in \mathbb{R}^{N_{\text{FCI}} \times N_{\text{FCI}}}, \quad \mathbf{c} \in \mathbb{R}^{N_{\text{FCI}}}$$

# FCI Matrix Properties

Entry:  $H_{ij} = \langle D_i | \hat{H} | D_j \rangle$ , not guaranteed to be non-negative.

- Symmetric:  $H_{ij} = H_{ji}$ .
- Sparse:  $H_{ij} = 0$  if  $|D_i\rangle$  and  $|D_j\rangle$  differ by more than two orbitals.  $H$  has  $O(n_{\text{elec}}^2 n_{\text{orb}}^2)$  entries per row.
- Ground-state eigenvalue  $E_0 < 0$ , ground-state eigenvector  $\mathbf{v}_0$  sparse in the sense of truncation.

# Memory usage

Table: Different Molecule Systems and Storage cost

Molecule	Basis	Electrons	Spin-Orbitals	Dimension	Memory
H <sub>2</sub> O	cc-pVDZ	10	48	$\sim 10^8$	$\sim 1$ GB
N <sub>2</sub>	cc-pVDZ	14	56	$\sim 10^{11}$	$\sim 1$ TB
N <sub>2</sub>	cc-pVTZ	14	120	$\sim 10^{16}$	$\sim 100$ PB
Cr <sub>2</sub>	Ahlrichs	48	84	$\sim 10^{22}$	-

Solution: Select a part of configurations (on-the-fly)!



# Basic Algorithmic Steps of Selected CI Program

- 1 Determine an initial variational space.
- 2 Find the ground state of the Hamiltonian in the current variational space.
- 3 Which determinants outside the space are most important? Perform some search algorithm based on perturbative or energy minimization estimates.
- 4 Construct a new variational space based on the search results.
- 5 Go to step 3, continue until the space stop changing.

# Notations

notation	description
$\hat{H}$	Hamiltonian operator
$H$	Hamiltonian matrix
$H_{ij}$	the $i, j$ -th entry of the Hamiltonian matrix
$E_0$	ground-state energy
$ \Phi_0\rangle$	ground-state wavefunction
$ D_i\rangle$	the $i$ -th Slater determinant
$\mathbf{c}$	wavefunction coefficient vector
$\mathbf{b}$	$\mathbf{b} = H\mathbf{c}$

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# Perturbation Theory

If we have ground-state wavefunction  $|\Phi_0\rangle = \sum_{i=1} c_i |D_i\rangle$ , then

$$c_i = \frac{\sum_{j \neq i} H_{ij} c_j}{H_{ii} - E_0}.$$

# Selecting Important Determinants

Different methods define different importance measure  $f(D_k)$ .

- Configuration Interaction by Perturbatively Selecting Iteration (CIPSI) or Adaptive Sampling Configuration Interaction (ASCI):

$$f_{\text{CIPSI}}(D_k) = \left| \frac{\sum_i H_{ki} c_i}{H_{kk} - E_0} \right|.$$

- Heat-Bath Configuration Interaction (HCI):

$$f_{\text{HCI}}(D_k) = \max_i (|H_{ki} c_i|),$$

where  $D_i$  in selected space.

# Coordinate Descent FCI (CDFCI)

FCI eigenvalue problem to unconstrained minimization problem

$$\min_{\mathbf{c} \in \mathbb{R}^{N_{\text{FCI}}}} f(\mathbf{c}) = \min_{\mathbf{c}} \|H + \mathbf{c}\mathbf{c}^T\|_F^2$$

- The only two local minimizers are  $\pm\sqrt{E_0}\mathbf{v}_0$ .
- Ensures convergence to the ground state wavefunction, given a good starting point (Hartree–Fock ground state).

## Coordinate gradient descent method

- Minimizes computational costs by **avoiding operations with the entire Hamiltonian matrix**.
- In each iteration, only *one coordinate* of the optimizing vector is updated.
- Computation for updating involves only *one column* of the Hamiltonian matrix.

# CDFCI Framework

Store  $\mathbf{c}$  and  $\mathbf{b} = H\mathbf{c}$  in memory. Initialize  $\mathbf{c}^{(0)}$ ,  $\mathbf{b}^{(0)} = H\mathbf{c}^{(0)}$ .

For iteration  $\ell = 1, 2, \dots$

- ① Select coordinate  
 $i^{(\ell)} = \arg \max_i |\nabla_i f(\mathbf{c}^{(\ell-1)})|$ .
- ② Find stepsize by exact line search  
 $\alpha^{(\ell)} = \arg \min_{\alpha} f(\mathbf{c}^{(\ell-1)} + \alpha \mathbf{e}_{i^{(\ell)}})$ .
- ③ Update  $\mathbf{c}^{(\ell)} = \mathbf{c}^{(\ell-1)} + \alpha^{(\ell)} \mathbf{e}_{i^{(\ell)}}$ ,  
 $\mathbf{b}^{(\ell)} = \mathbf{b}^{(\ell-1)} + \alpha^{(\ell)} H_{:,i^{(\ell)}}$ .

Remark: Gradient

$$\nabla f(\mathbf{c}) = 4H\mathbf{c} + 4\mathbf{c}^T \mathbf{c} \mathbf{c} = 4\mathbf{b} + 4\mathbf{c}^T \mathbf{c} \mathbf{c}.$$

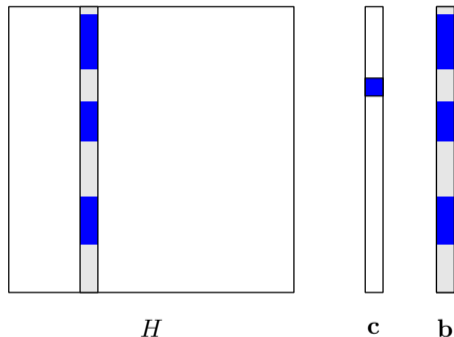


Figure: Update for one coordinate.

# CDFCI Framework - for Two Coordinates?

Initialize  $\mathbf{c}^{(0)}$ ,  $\mathbf{b}^{(0)} = H\mathbf{c}^{(0)}$ .

For iteration  $\ell = 1, 2, \dots$

- 1 Select coordinate  
 $i^{(\ell)} = \arg \max_i |\nabla_i f(\mathbf{c}^{(\ell-1)})|$ ,  
 $j^{(\ell)} = \arg \max_{j \neq i^{(\ell)}} |\nabla_j f(\mathbf{c}^{(\ell-1)})|$ .
- 2 Find stepsize  
 $\alpha^{(\ell)} = \arg \min_{\alpha} f(\mathbf{c}^{(\ell-1)} + \alpha \mathbf{e}_{i^{(\ell)}})$ ,  
 $\beta^{(\ell)} = \arg \min_{\beta} f(\mathbf{c}^{(\ell-1)} + \beta \mathbf{e}_{j^{(\ell)}})$ .
- 3 Update  $\mathbf{c}^{(\ell)} = \mathbf{c}^{(\ell-1)} + \alpha^{(\ell)} \mathbf{e}_{i^{(\ell)}} + \beta^{(\ell)} \mathbf{e}_{j^{(\ell)}}$ ,  
 $\mathbf{b}^{(\ell)} = \mathbf{b}^{(\ell-1)} + \alpha^{(\ell)} H_{:,i^{(\ell)}} + \beta^{(\ell)} H_{:,j^{(\ell)}}$ .

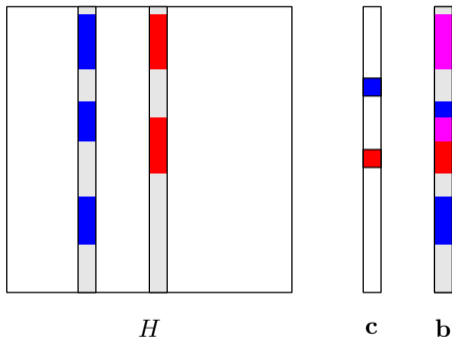


Figure: Update for two coordinates.



# CDFCI Framework - Exact Line Search?

Initialize  $\mathbf{c}^{(0)}$ ,  $\mathbf{b}^{(0)} = H\mathbf{c}^{(0)}$ .

For iteration  $\ell = 1, 2, \dots$

- 1 Select coordinate

$$i^{(\ell)} = \arg \max_i |\nabla_i f(\mathbf{c}^{(\ell-1)})|,$$

$$j^{(\ell)} = \arg \max_{j \neq i^{(\ell)}} |\nabla_j f(\mathbf{c}^{(\ell-1)})|.$$

- 2 Find stepsize  $\alpha^{(\ell)}, \beta^{(\ell)} =$

$$\arg \min_{\alpha, \beta} f(\mathbf{c}^{(\ell-1)} + \alpha \mathbf{e}_{i^{(\ell)}} + \beta \mathbf{e}_{j^{(\ell)}}).$$

- 3 Update  $\mathbf{c}^{(\ell)} = \mathbf{c}^{(\ell-1)} + \alpha^{(\ell)} \mathbf{e}_{i^{(\ell)}} + \beta^{(\ell)} \mathbf{e}_{j^{(\ell)}}$ ,

$$\mathbf{b}^{(\ell)} = \mathbf{b}^{(\ell-1)} + \alpha^{(\ell)} H_{:,i^{(\ell)}} + \beta^{(\ell)} H_{:,j^{(\ell)}}.$$

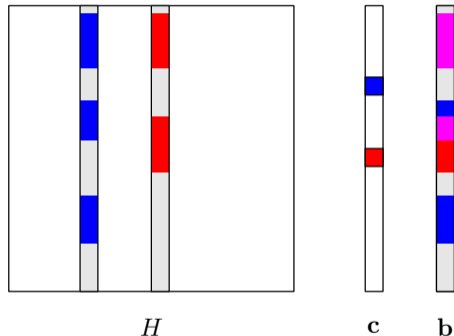


Figure: Update for two coordinates.

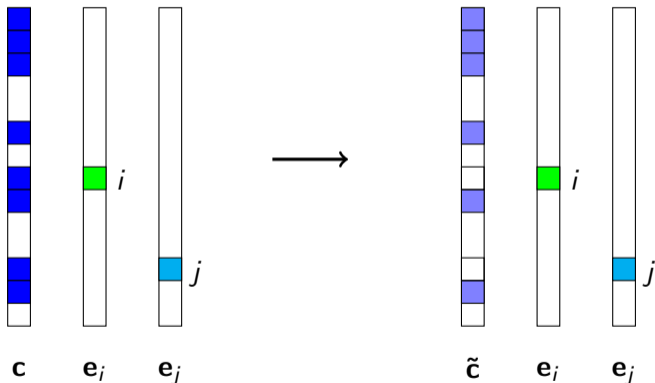
# Add a Scalar $\gamma$ for Exact Line Search

Modify the minimization problem from  $\min_{\alpha, \beta} f(\mathbf{c} + \alpha \mathbf{e}_i + \beta \mathbf{e}_j)$  to

$$\begin{aligned} \min_{\gamma, \alpha, \beta} f(\gamma \mathbf{c} + \alpha \mathbf{e}_i + \beta \mathbf{e}_j) &= f\left(\begin{bmatrix} \mathbf{c} & \mathbf{e}_i & \mathbf{e}_j \end{bmatrix} \begin{bmatrix} \gamma \\ \alpha \\ \beta \end{bmatrix}\right) \\ &= \left\| H + \begin{bmatrix} \mathbf{c} & \mathbf{e}_i & \mathbf{e}_j \end{bmatrix} \begin{bmatrix} \gamma \\ \alpha \\ \beta \end{bmatrix} \begin{bmatrix} \gamma & \alpha & \beta \end{bmatrix} \begin{bmatrix} \mathbf{c}^T \\ \mathbf{e}_i^T \\ \mathbf{e}_j^T \end{bmatrix} \right\|_F^2. \end{aligned}$$

# Matrix Orthogonalization

Construct  $[\tilde{\mathbf{c}} \ \mathbf{e}_i \ \mathbf{e}_j]$ , where  $\|\tilde{\mathbf{c}}\|_2 = 1$ ,  $(\tilde{\mathbf{c}}, \mathbf{e}_i) = 0$ ,  $(\tilde{\mathbf{c}}, \mathbf{e}_j) = 0$ .



# Add $\gamma$ and $\tilde{\mathbf{c}}$ for Exact Line Search

Modify the minimization problem from  $\min_{\alpha, \beta} f(\mathbf{c} + \alpha \mathbf{e}_i + \beta \mathbf{e}_j)$  to

$$\begin{aligned}
 \min_{\gamma, \alpha, \beta} f(\gamma \tilde{\mathbf{c}} + \alpha \mathbf{e}_i + \beta \mathbf{e}_j) &= f\left(\begin{bmatrix} \tilde{\mathbf{c}} & \mathbf{e}_i & \mathbf{e}_j \end{bmatrix} \begin{bmatrix} \gamma \\ \alpha \\ \beta \end{bmatrix}\right) \\
 &= \left\| \left( H + \begin{bmatrix} \tilde{\mathbf{c}} & \mathbf{e}_i & \mathbf{e}_j \end{bmatrix} \begin{bmatrix} \gamma \\ \alpha \\ \beta \end{bmatrix} \begin{bmatrix} \gamma & \alpha & \beta \end{bmatrix} \begin{bmatrix} \tilde{\mathbf{c}}^T \\ \mathbf{e}_i^T \\ \mathbf{e}_j^T \end{bmatrix} \right) \right\|_F^2 \\
 &= \left\| \underbrace{\begin{bmatrix} \tilde{\mathbf{c}}^T \\ \mathbf{e}_i^T \\ \mathbf{e}_j^T \end{bmatrix} H \begin{bmatrix} \tilde{\mathbf{c}} & \mathbf{e}_i & \mathbf{e}_j \end{bmatrix}}_{\in \mathbb{R}^{3 \times 3}} + \begin{bmatrix} \gamma \\ \alpha \\ \beta \end{bmatrix} \begin{bmatrix} \gamma & \alpha & \beta \end{bmatrix} \right\|_F^2.
 \end{aligned}$$

# Extension to Multi Coordinate Descent FCI

- Select a set of coordinates  $I = \{i_1, \dots, i_k\}$ ,  $1 \leq i_j \leq N_{\text{FCI}}$  based on gradient  $\nabla f(\mathbf{c}) = 4H\mathbf{c} + 4\mathbf{c}^T\mathbf{c}\mathbf{c}$ .
- Denote  $\mathcal{E}_I = [e_{i_1}, \dots, e_{i_k}] \in \mathbb{R}^{N_{\text{FCI}} \times k}$ .
- The update is given by

$$\mathbf{c} \leftarrow \gamma\mathbf{c} + \mathcal{E}_I\mathbf{a}.$$

- The values of  $\gamma$  and  $\mathbf{a}$  are given by the eigenvector of

$$\begin{bmatrix} \tilde{\mathbf{c}}^T \\ \mathcal{E}_I^T \end{bmatrix} H \begin{bmatrix} \tilde{\mathbf{c}} & \mathcal{E}_I \end{bmatrix} \in \mathbb{R}^{(k+1) \times (k+1)}$$

corresponding to the minimal eigenvalue  $\lambda_{\min}$ , which is the current energy estimate.

# Epstein–Nesbet(EN) Energy

Adding  $E_{\text{EN}}$  to  $E_{\text{var}}$  to recover the energy outside variational space.

$$E_{\text{EN}} = \sum_k \frac{(\sum_i H_{ki} c_i)^2}{E_{\text{var}} - H_{kk}},$$

where  $k$  iterates over determinants excluded.

To accelerate,

$$E_{\text{EN}} \approx \sum_k \frac{(\sum_i^{|H_{ki} c_i| > \epsilon} H_{ki} c_i)^2}{E_{\text{var}} - H_{kk}}.$$

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# Parallel ASCI

- Classify determinants by their three highest-occupied orbitals
- Pre-compute the amount of work for each set
- Assign work to individual processing units

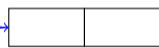
$$D_j \in |\Phi_0\rangle: D_j \rightarrow \tilde{D}_i \in \{D_j^{\text{sd}}\}$$

$$\text{Compute } S_i^{(j)} = \frac{H_{ij}C_j}{H_{ii} - E_{\text{var}}}$$



$$(\tilde{D}_i, S_i^{(j)})$$

Aggregate on  $\tilde{D}_i$



$$(\tilde{D}_i, S_i = \sum_j S_i^{(j)})$$

Sort on  $S_i$



## Parallel HCI

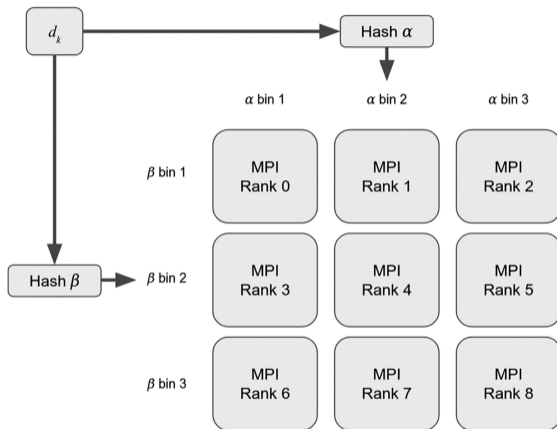
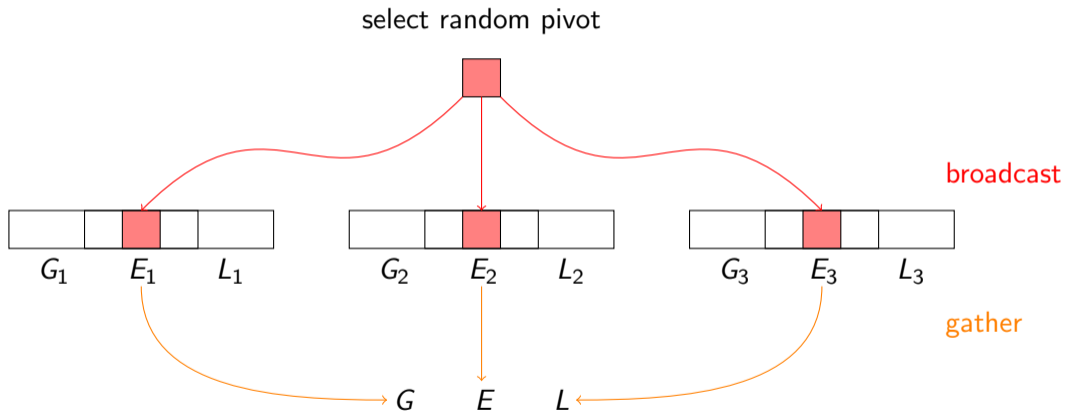


Figure: 2D hash over  $\alpha$  and  $\beta$  strings to assign MPI Ranks

# QuickSelect instead of QuickSort



# Summary

- Configuration Interaction Method transforms the many-body time-independent Schrödinger equation to FCI eigenvalue problem under a basis consisting of Slater determinants of configuration combinations.
- Full CI is exact (under predefined one-electron basis) but too expensive; selected CI + perturbation provides approximations.
- Coordinate Descent FCI can be categorized as one of the SCI+PT methods.
- It can be extended to a multi-coordinate version with higher scalability.

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*Thanks for Your Attention!*