

# Parallel Coordinate Descent Methods for Full Configuration Interaction

Yuejia Zhang   Weiguo Gao   Yingzhou Li

Fudan University, China

SIAM LA 24, Paris, France

# Problem Statement

- Search for the **ground-state** of a chemical system given by the **many-body** time-independent Schrödinger Equation

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

where  $|\Phi_0\rangle = \Phi_0(r_1, \dots, r_{n_{\text{elec}}})$ ,  $r_i \in \mathbb{R}^3$ .

- Under Born–Oppenheimer approximation, the Hamiltonian operator with  $n_{\text{nuc}}$  nuclei and  $n_{\text{elec}}$  electrons is

$$\hat{H} = -\frac{1}{2} \sum_{i=1}^{n_{\text{elec}}} \nabla_i^2 + \sum_{i=1}^{n_{\text{elec}}} V_{\text{ext}}(r_i; \{R_I\}_{I=1}^{n_{\text{nuc}}}) + \sum_{i < j}^{n_{\text{elec}}} \frac{1}{\|r_i - r_j\|}.$$

# FCI Numerical Discretization

- 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 FCI eigenvalue problem

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

# Hamiltonian Matrix

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

- Symmetric.  $H_{ij} = H_{ji}$ .
- Sparse. For off-diagonals  $|D_i\rangle \neq |D_j\rangle$ ,
  - If  $|D_i\rangle = a_r^\dagger a_p |D_j\rangle$ ,  $H_{ij} = \langle r | \hat{h} | p \rangle + \sum_k \langle rk | |pk\rangle$ .
  - If  $|D_i\rangle = a_r^\dagger a_s^\dagger a_p a_q |D_j\rangle$ ,  $H_{ij} = \langle rs | |pq\rangle$ .
  - Otherwise,  $H_{ij} = 0$ .

Consequence:  $H$  has  $O(n_{\text{elec}}^2 n_{\text{orb}}^2)$  entries per row.

- Ground-state eigenvalue  $E_0 < 0$ .
- Ground-state eigenvector  $\mathbf{c}$  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>	Ahlich	48	84	$\sim 10^{22}$	-

## Solution: Wavefunction Compression

- By tensor train: DMRG
- By sampling: FCIQMC, iFCIQMC, S-FCIQMC
- By selecting “important” configurations: HCI, SHCI, ASCI, **CDFCI**

# FCI eigenvalue problem

Consider the unconstrained minimization problem

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

- Gradient  $\nabla f(\mathbf{c}) = 4H\mathbf{c} + 4(\mathbf{c}^T\mathbf{c})\mathbf{c}$ .
- Hessian  $\nabla^2 f(\mathbf{c}) = 4H + 8\mathbf{c}\mathbf{c}^T + 4(\mathbf{c}^T\mathbf{c})I$ .
- **Non-convex** problem, with **unbounded** Lipschitz constant.
- Stationary points:  $0, \pm\sqrt{-\lambda_1}\mathbf{v}_1, \dots, \pm\sqrt{-\lambda_m}\mathbf{v}_m$  ( $\dots < \lambda_m < 0 < \lambda_{m+1} < \dots$ ).
- Only two **local minimizers**  $\pm\sqrt{-\lambda_1}\mathbf{v}_1$  (which are also **global minimizers**), the others are all strict saddle points.
- Ensures convergence to the ground state  $\pm\mathbf{c}$ , given a good starting point (e.g., Hartree–Fock ground state).

# Coordinate Descent FCI (CDFCI)<sup>1</sup>

## 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.

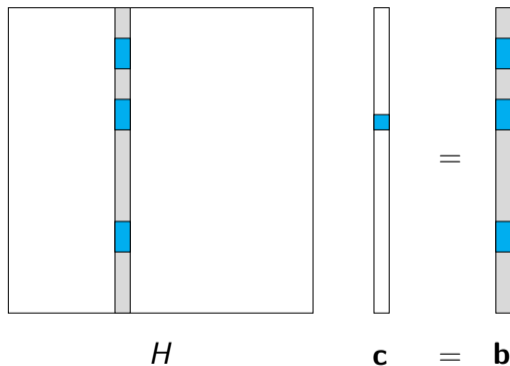


Figure: Update for one coordinate.

<sup>1</sup>Z. Wang, Y. Li, J. Lu, J. Chem. Theory Comput., 2019.

# CDFCI Framework

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

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

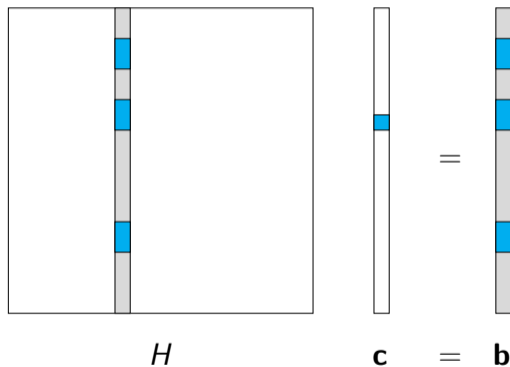


Figure: Update for one coordinate.



# 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

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

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

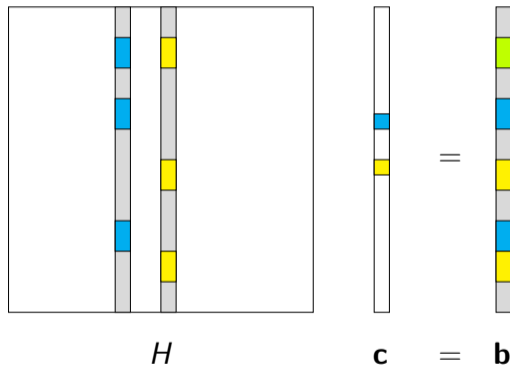


Figure: Update for two coordinates.

# 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

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

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

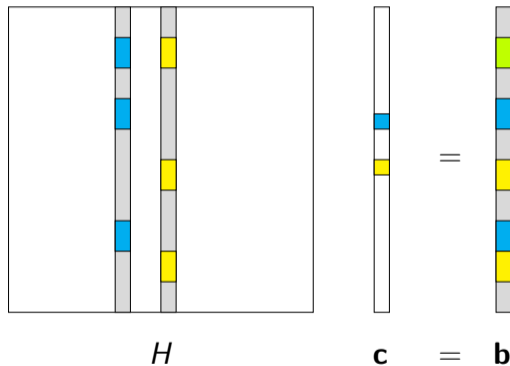


Figure: Update for two coordinates.

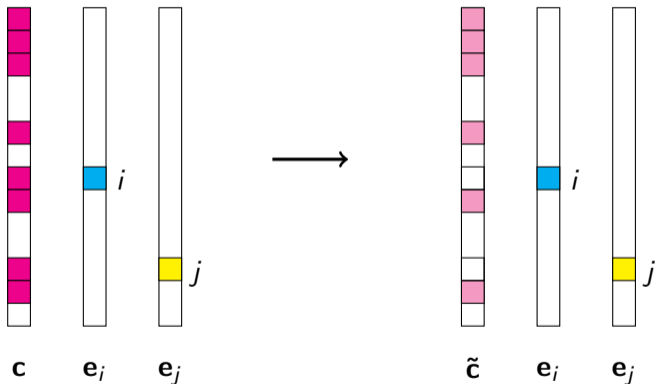
# Add a Scalar $\gamma$ for Exact Line Search

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

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

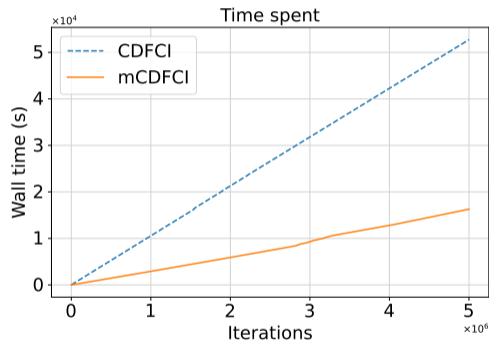
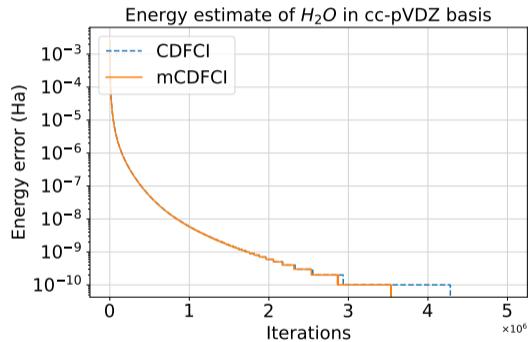
corresponding to the minimal eigenvalue  $\lambda_{\min}$ , which is the current energy estimate.

# Implementation Details

- $H_{ij}$  evaluated on-the-fly.
- $\mathbf{c}$  and  $\mathbf{b} = H\mathbf{c}$  are stored in a hash table.
- Update of  $\mathbf{b}_i$  is discarded if  $\mathbf{c}_i = 0$  and  $\Delta\mathbf{b}_i < \tau$ , where  $\Delta\mathbf{b} = H_{:j}a_j$ . Note that this does not affect eigenvalue estimator

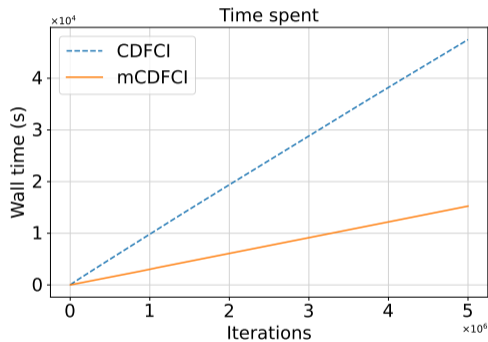
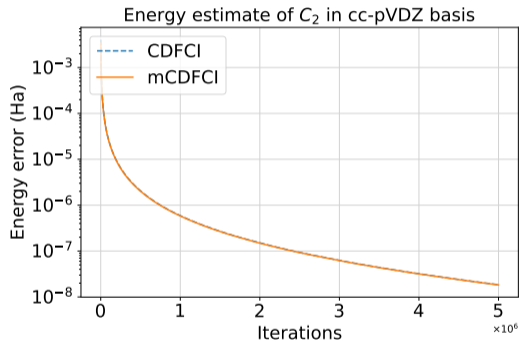
$$\text{RQ}(\mathbf{c}) = \frac{\mathbf{c}^T H \mathbf{c}}{\mathbf{c}^T \mathbf{c}} = \frac{\mathbf{c}^T \mathbf{b}}{\mathbf{c}^T \mathbf{c}}.$$

- Compression tolerance  $\tau$  balances between memory-cost and accuracy.
- Shared memory parallelism based on OpenMP: the updates of  $\mathbf{c}$  and  $\mathbf{b} = H\mathbf{c}$  for each coordinate are performed in parallel.

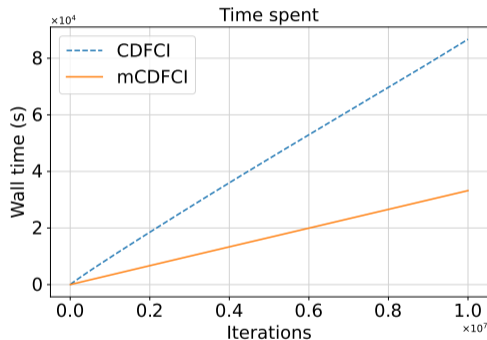
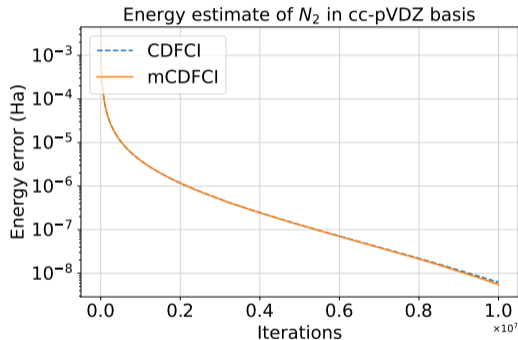
Overall Speedup:  $H_2O/cc\text{-pVDZ}$ 

**Figure:** Speedup of mCDFCI compared with CDFCI(2019), both in 64 threads. We perform  $k$  coordinates ( $k = 64$ ) descent per iteration for the original CDFCI.



Overall Speedup:  $C_2/cc\text{-pVDZ}$ 

**Figure:** Speedup of mCDFCI compared with CDFCI(2019), both in 64 threads. We perform  $k$  coordinates ( $k = 64$ ) descent per iteration for the original CDFCI.

Overall Speedup:  $N_2$ /cc-pVDZ

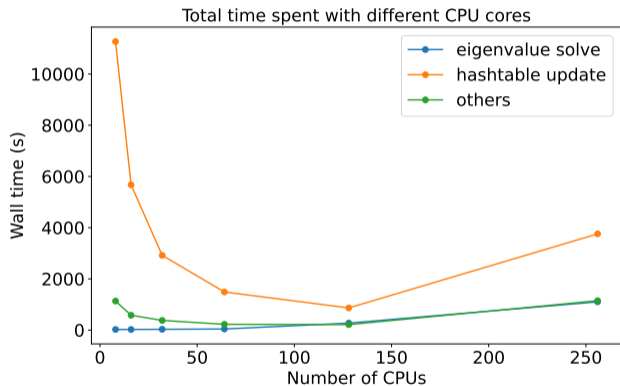
Remark: With a larger search space each step, mCDFCI leads to lower energy with just 0.2% more coordinates searched.

## Scalability

**Table:** Speedup of mCDFCI for different number of coordinates (threads) of H<sub>2</sub>O in cc-pVDZ basis

Energy	Error	Wall Time (s)	Speedup on $k$ Cores					
		on Single Core	2	4	8	16	32	64
-76.2318601	$10^{-2}$	9.0	1.9×	3.7×	5.0×	7.0×	10.4×	13.1×
-76.2408601	$10^{-3}$	292.9	2.0×	3.7×	4.8×	8.2×	14.3×	20.3×
-76.2417601	$10^{-4}$	1837.6	2.0×	3.5×	6.1×	8.3×	16.2×	22.5×
-76.2418501	$10^{-5}$	9016.7	2.1×	4.1×	8.0×	12.3×	21.1×	29.0×
-76.2418591	$10^{-6}$	32931.7	2.1×	4.5×	9.3×	16.2×	29.1×	40.7×

# Scalability for each procedure



**Figure:** Time spent listed in each procedure, while running  $6.4\text{M core} \times \text{iterations}$  for  $Cr_2$  in Ahlrics SV Basis ( $48e, 84o$ ),  $\tau = 10^{-4}$ .

# Summary

The proposed methods **CDFCI** and **mCDFCI**

- performs configuration selection using **coordinate descent** and **exact line search**.
- visits important determinants efficiently.
- captures the significant part of FCI space for ground state approximation.

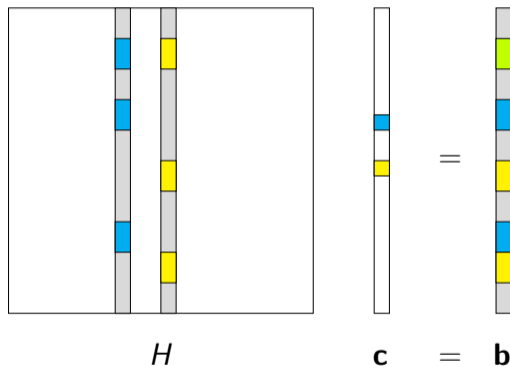


Figure: Update for two coordinates.

# *Thanks for Your Attention!*

Registration and travel support for this presentation was provided by the Society for Industrial and Applied Mathematics.