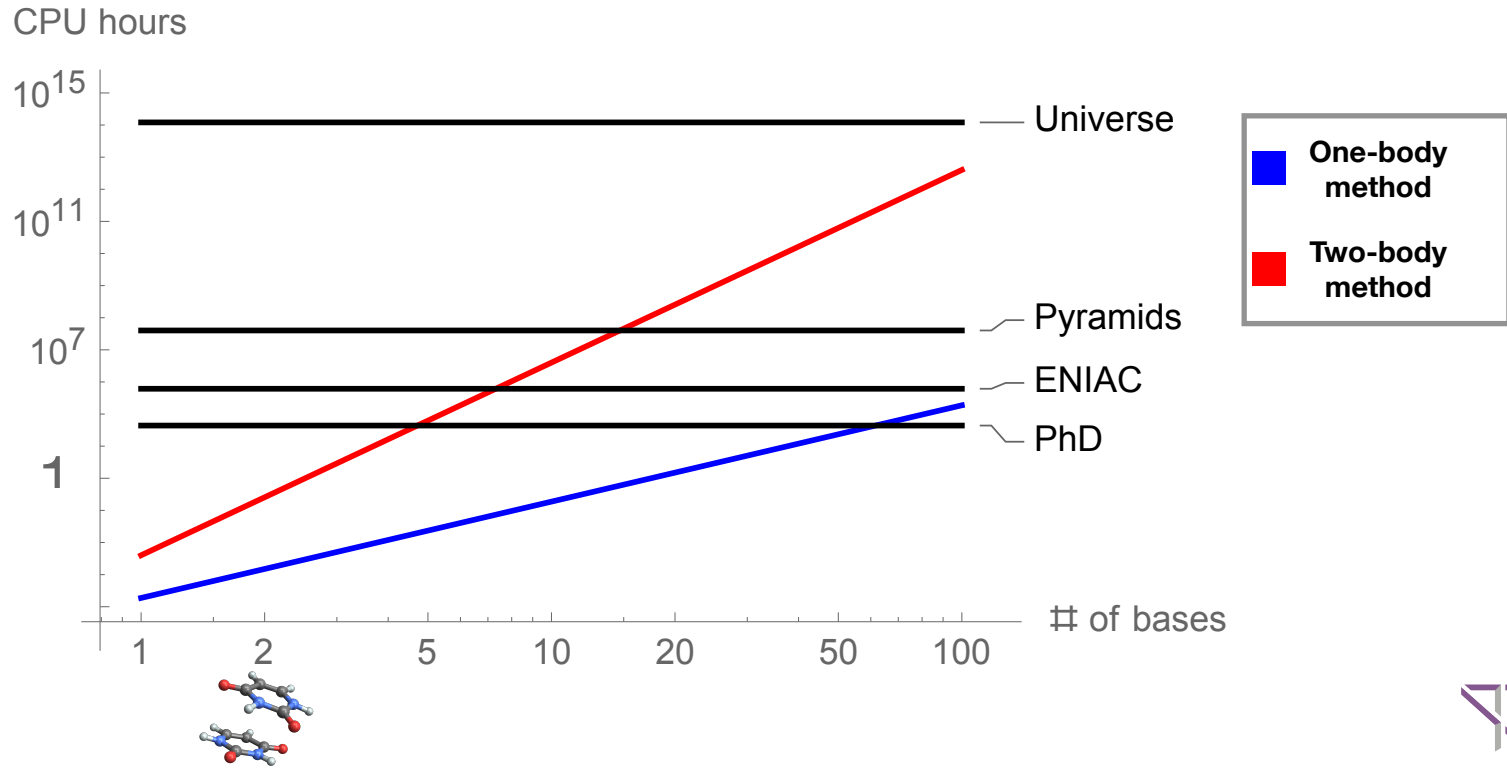


APPROXIMATING TENSOR CONTRACTIONS VIA A MATRIX-FREE TENSOR DECOMPOSITION

KARL PIERCE

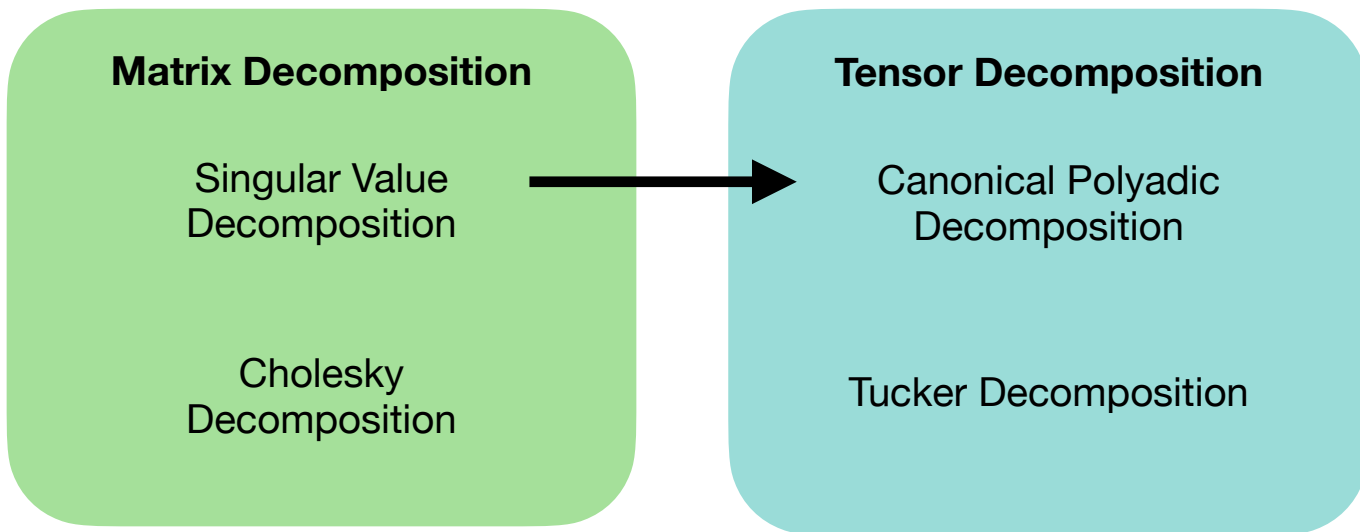
The cost of molecular electronic structure methods



The correlation between storage and computational cost

	Computational Scaling	Storage	Element of highest order tensor
CCSD	$\mathcal{O}(N^6)$	$\mathcal{O}(N^4)$	t_{ab}^{ij}
CCSDT	$\mathcal{O}(N^8)$	$\mathcal{O}(N^6)$	t_{abc}^{ijk}
CCSD(T)	$\mathcal{O}(N^7)$	$\mathcal{O}(N^4)$	t_{ab}^{ij}

Tensor compression techniques



Singular value decomposition (SVD)

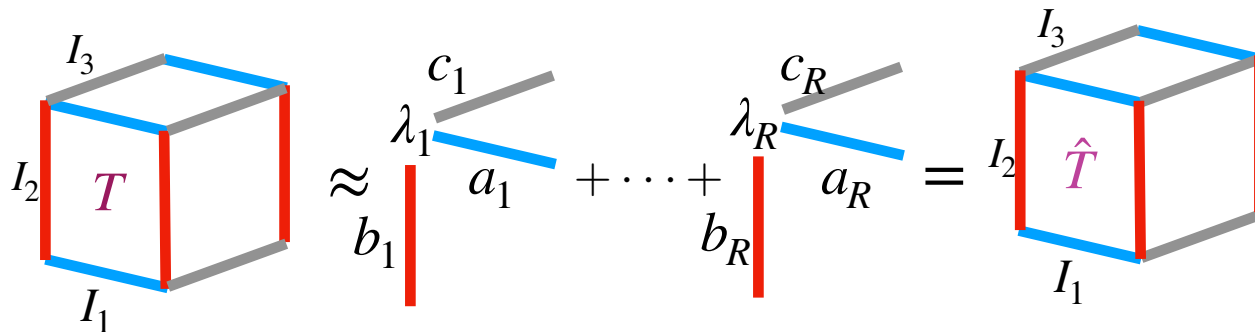
$$\begin{array}{c} I_1 \\ I_2 \end{array} \begin{array}{|c|} \hline T \\ \hline \end{array} = \begin{array}{c} \lambda_1 \\ \text{red bar} \\ b_1 \end{array} \overbrace{a_1}^{\text{blue bar}} + \cdots + \begin{array}{c} \lambda_R \\ \text{red bar} \\ b_R \end{array} \overbrace{a_R}^{\text{blue bar}}$$

$$\begin{array}{c} I_1 \\ I_2 \end{array} \begin{array}{|c|} \hline T \\ \hline \end{array} = \begin{array}{c} I_1 \\ I_2 \end{array} \begin{array}{|c|} \hline \text{red box} \\ \hline \end{array} + \cdots + \begin{array}{c} I_1 \\ I_2 \end{array} \begin{array}{|c|} \hline \text{pink box} \\ \hline \end{array}$$

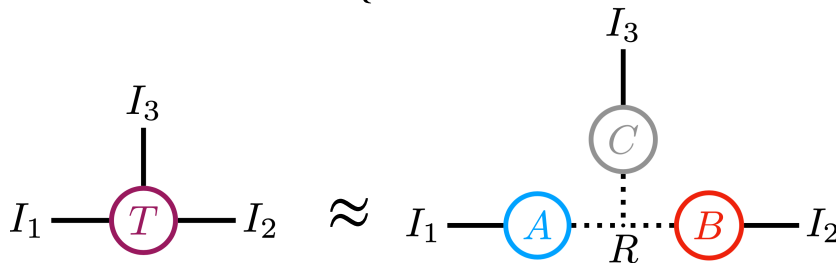
$$I_1 \text{---} \textcircled{T} \text{---} I_2 = I_1 \text{---} \textcircled{A} \cdots \textcircled{B} \text{---} I_2$$

R

Canonical Polyadic Decomposition (CPD)



$$A = \{a_r\} \in \mathcal{R}^{I_1 \times R}$$



$$\mathcal{O}(I^3) \rightarrow \mathcal{O}(3IR)$$

Coupled Cluster Optimization

Compute the coupled cluster wavefunction

$$|\Psi_{cc}\rangle = e^{\hat{T}} |\Psi_0\rangle$$

$$\hat{T} = \sum_{ia} t_i^a a_a^i + \sum_{ijab} t_{ij}^{ab} a_{ab}^{ij} + \dots$$

Truncate the cluster operator at n th-excitation

Coupled Cluster Optimization

Compute the coupled cluster wavefunction

$$|\Psi_{cc}\rangle = e^{\hat{T}} |\Psi_0\rangle$$

$$\hat{T}_{ccsd} = \sum_{ia} t_i^a a_a^i + \sum_{ijab} \boxed{t_{ij}^{ab}} a_{ab}^{ij}$$

Goal: Optimize the n excitation tensors

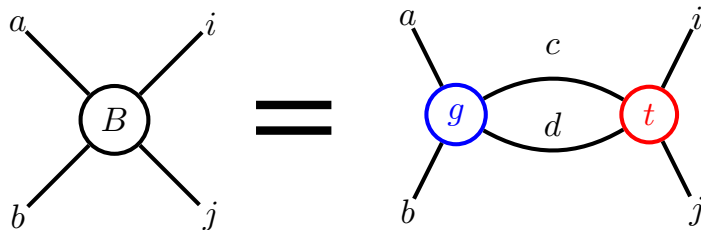
CCSD Optimization

t_{ij}^{ab} tensor optimization involves

$${}^nR_{ij}^{ab} = A_{ij}^{ab} + B_{ij}^{ab} + C_{ij}^{ab} + D_{ij}^{ab} \qquad {}^n t = {}^{n-1} t + \alpha({}^n R)$$

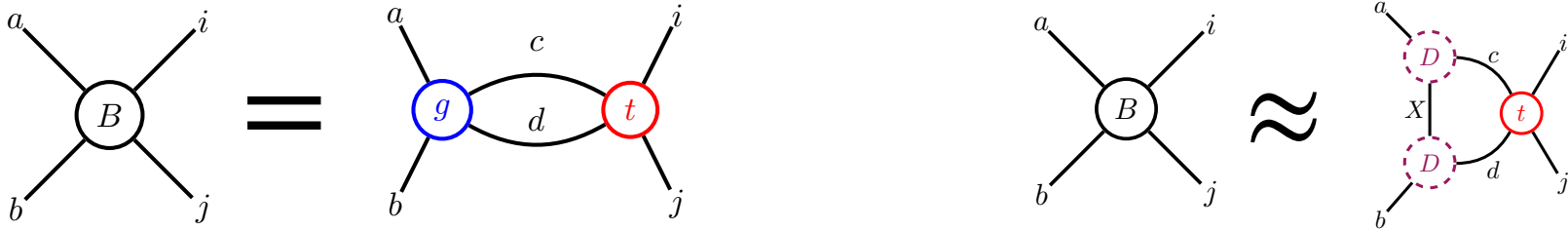
Particle-Particle Ladder Diagram

$$B_{ij}^{ab} = \sum_{cd} g_{cd}^{ab} t_{ij}^{cd}$$



It is standard to approximate tensors in a network

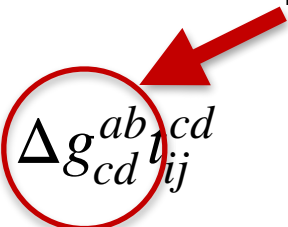
$$B_{ij}^{ab} = \sum_{cd} g_{cd}^{ab} t_{ij}^{cd} \xrightarrow{\text{Density Fitting Approximation}} B_{ij}^{ab} \approx \sum_{cdX} D_c^{aX} D_d^{bX} t_{ij}^{cd}$$

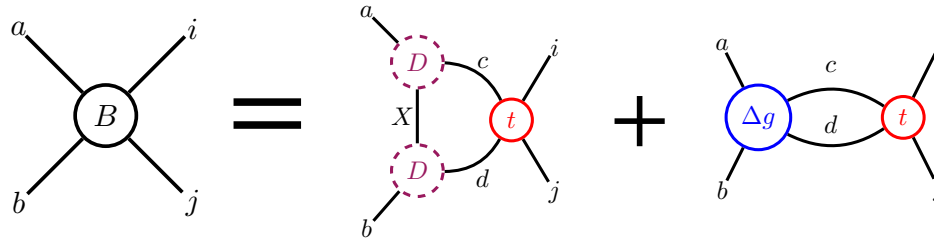


Error associated with approximated tensors in a network is amplified.

$$B_{ij}^{ab} = \sum_{cdX} D_c^{aX} D_d^{bX} t_{ij}^{cd} + \sum_{cdX} \Delta g_{cd}^{ab} t_{ij}^{cd}$$


DF approximation error





Error associated with approximated tensors in a network is amplified.

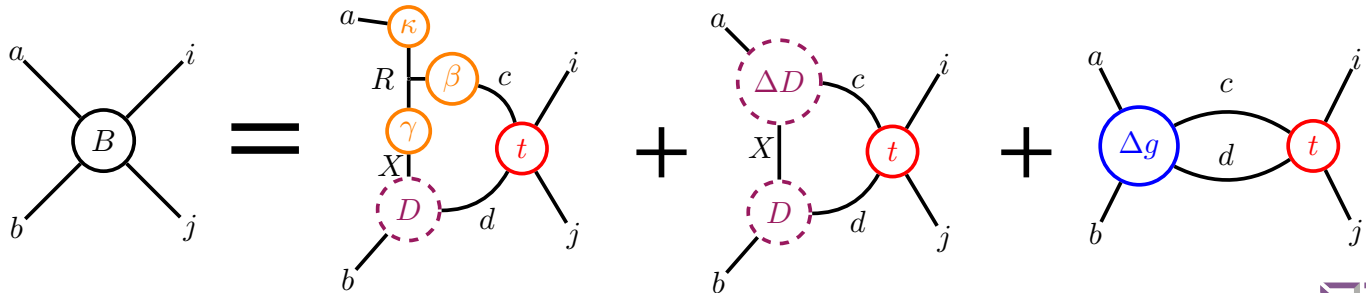
CPD approximation

$$\sum_{cd} g_{cd}^{ab} t_{ij}^{cd} \approx \sum_{cd} \sum_X \hat{D}_c^{aX} D_d^{bX} t_{ij}^{cd}$$


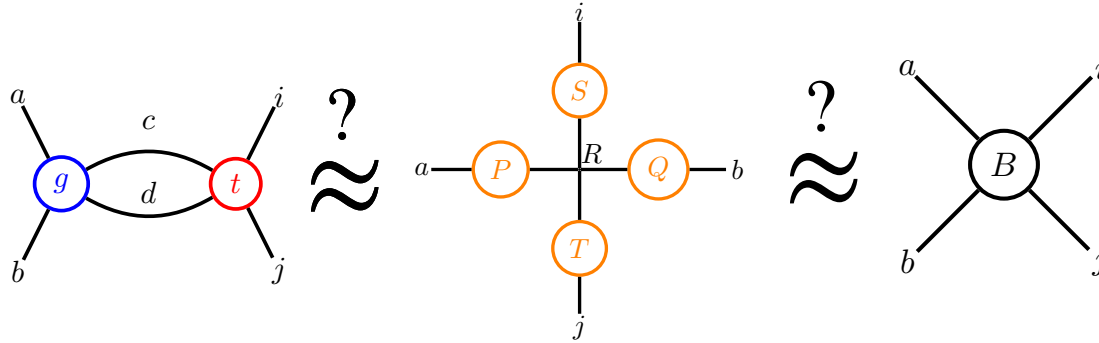
Error associated with approximated tensors in a network is amplified.

CPD approximation
error

$$\sum_{cd} g_{cd}^{ab} t_{ij}^{cd} = \sum_{cd} \left(\sum_X \hat{D}_c^{aX} D_d^{bX} + \sum_X \Delta D_c^{aX} D_d^{bX} + \Delta g_{cd}^{ab} \right) t_{ij}^{cd}$$



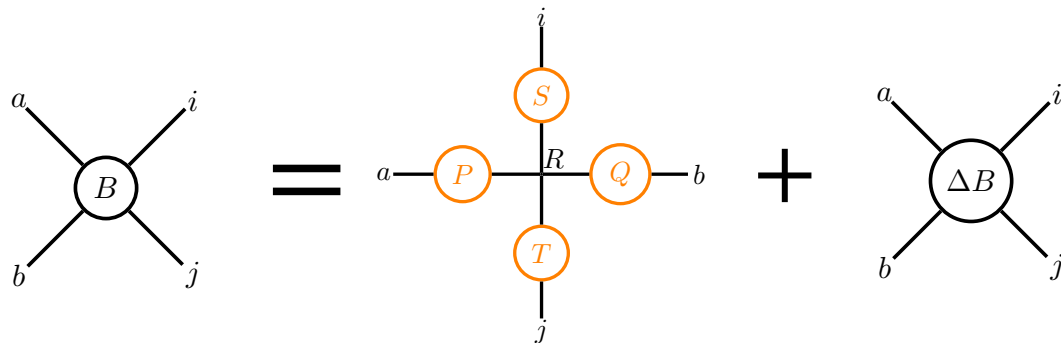
Is it possible to approximate the tensor contraction as a network?



$$f(x) = \frac{1}{2} \left\| \sum_{cd} g_{cd}^{ab} t_{ij}^{cd} - \hat{B}_{ij}^{ab}(x) \right\|_2^2$$

$$\hat{B}_{ij}^{ab}(x) = \sum_r^{R_{cp}} P_r^a Q_r^b S_i^r T_j^r$$

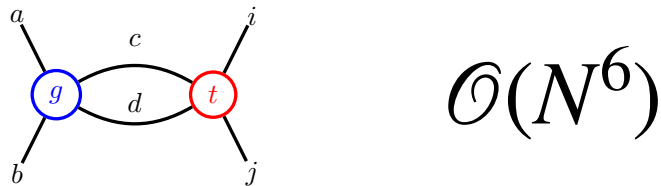
Is it possible to approximate the tensor contraction as a network?



$$B_{ij}^{ab} = \hat{B}_{ij}^{ab} + \Delta_{CPD} B_{ij}^{ab}$$

$\Delta_{CPD} B$ is directly controllable via the CPD

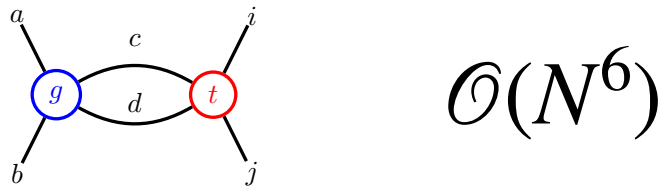
Side note: the CPD optimization of B is cheaper than the canonical contraction



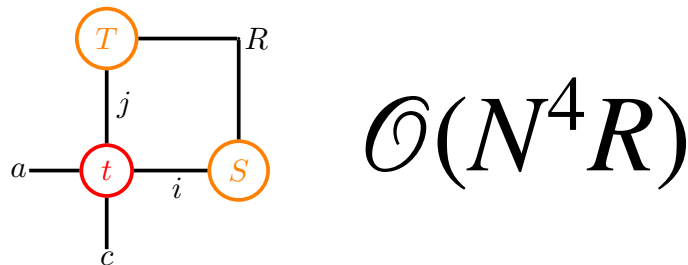
$$\frac{\partial f(x)}{\partial P} = \sum_{cd} \left(\sum_b g_{cd}^{ab} Q_r^b \right) \left(\sum_{ij} t_{ij}^{cd} S_i^r T_j^r \right) - P_r^a (Q_r^b Q_{r'}^b) * (S_i^r S_i^{r'}) * (T_j^r T_j^{r'})$$



Side note: the CPD optimization of B is cheaper than the canonical contraction



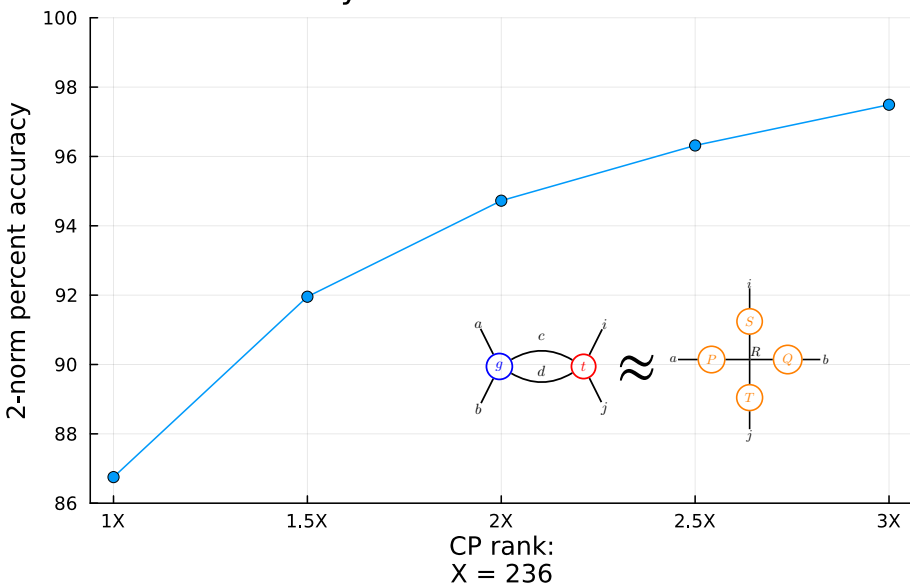
$$\frac{\partial f(x)}{\partial P} = \sum_{cd} g_{cd}^{ar} \left(\sum_{ij} t_{ij}^{cd} S_i^r T_j^{r'} \right) - P_r^a (Q_r^b Q_{r'}^b) * (S_i^r S_i^{r'}) * (T_j^r T_j^{r'})$$



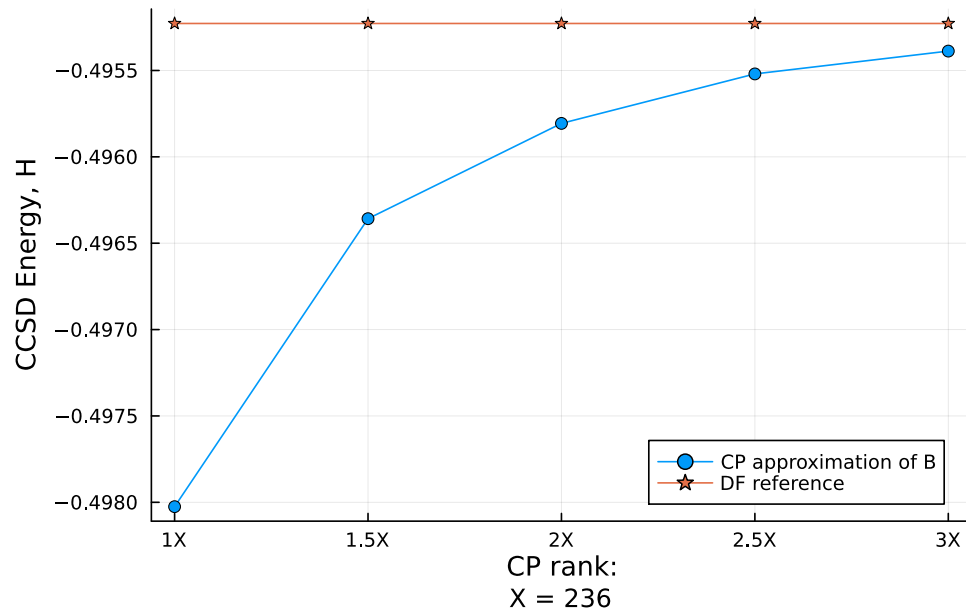
CP approximation of PPL tensor network

$(H_2O)_2$ TIP4P geometry
DZ-F12/aug-DZ-RI

Accuracy of the CPD of the B tensor



CPD approximated CCSD Energy



18 percent accuracy = $(1.0 - \|B - \hat{B}\|_2 / \|B\|_2) * 100$

How does one choose a stopping condition?

Canonical stopping condition:
2-norm error

$$\Delta_n = \|B_{ij}^{ab} - \hat{B}_{ij}^{ab}(x_n)\|$$

$$\Delta_n - \Delta_{n-1} < \epsilon_{ALS}$$

Computational complexity: $\mathcal{O}(N^6)$

Proposed stopping condition:
2-norm error

$$\Upsilon_n = \|\hat{B}_{ij}^{ab}(x_n) - \hat{B}_{ij}^{ab}(x_{n-1})\|$$

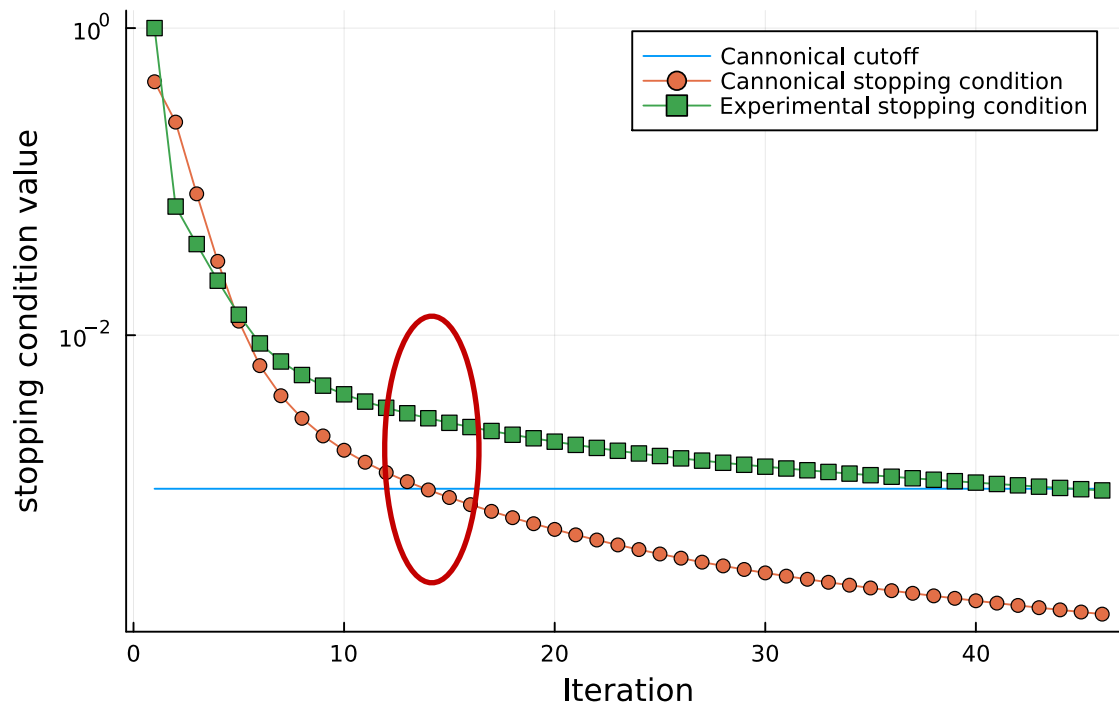
$$\Upsilon_n < \epsilon_{ALS}$$

Computational complexity: $\mathcal{O}(N^3)$

Comparing stopping conditions

$(H_2O)_2$ TIP4P geometry
DZ-F12/aug-DZ-RI

Convergence of stopping conditions



Result

It is feasible to approximate a tensor network contraction as a matrix free tensor decomposition

Question: Can we make this CPD
“contraction” more efficient?

Is there a way to reduce the cost of the CPD?

$${}^nR_{ij}^{ab} = A_{ij}^{ab} + B_{ij}^{ab} + C_{ij}^{ab} + D_{ij}^{ab}$$

$${}^nt = {}^{n-1}t + \alpha({}^nR)$$

Iteration 1: ${}^1B_{ij}^{ab} = g_{cd}^{ab}({}^1t_{ij}^{cd})$

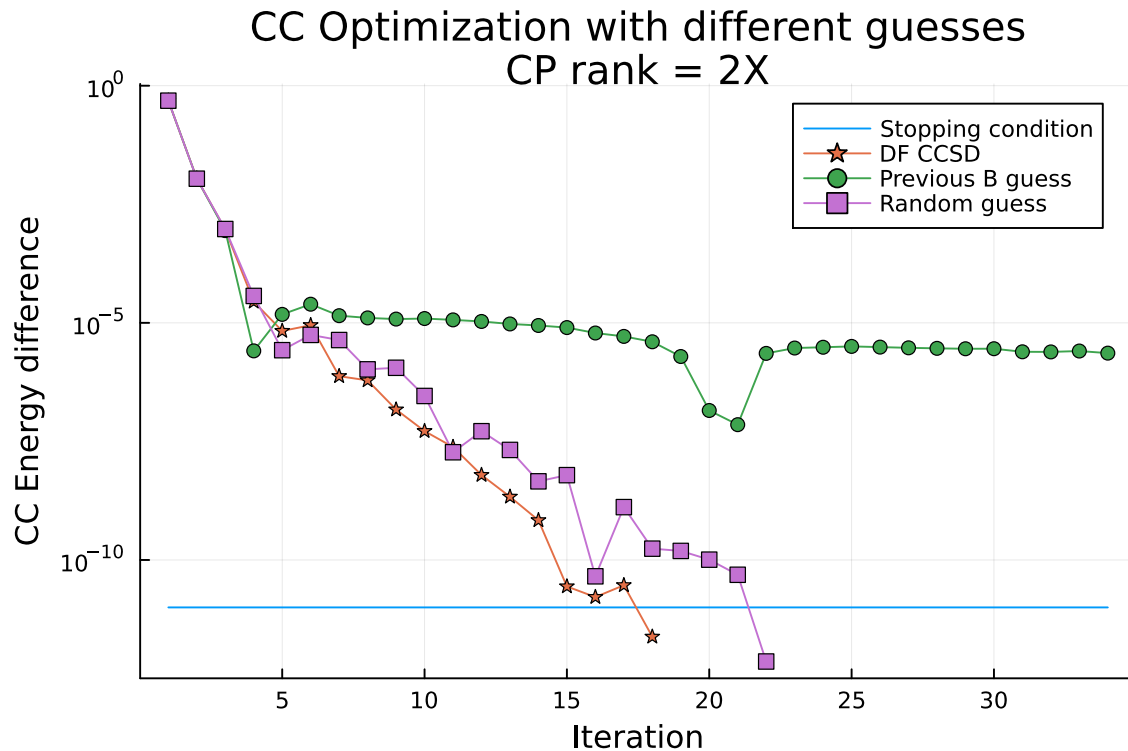
Iteration 2: ${}^2B_{ij}^{ab} = g_{cd}^{ab}({}^2t_{ij}^{cd})$ ${}^2B_{ij}^{ab} \approx g_{cd}^{ab}({}^1t_{ij}^{cd} + \delta^1 t_{ij}^{cd})$

Idea:

The optimized ${}^{n-1}\hat{B}$ factors might be a good initial guess
for the n th CC iteration's CPD

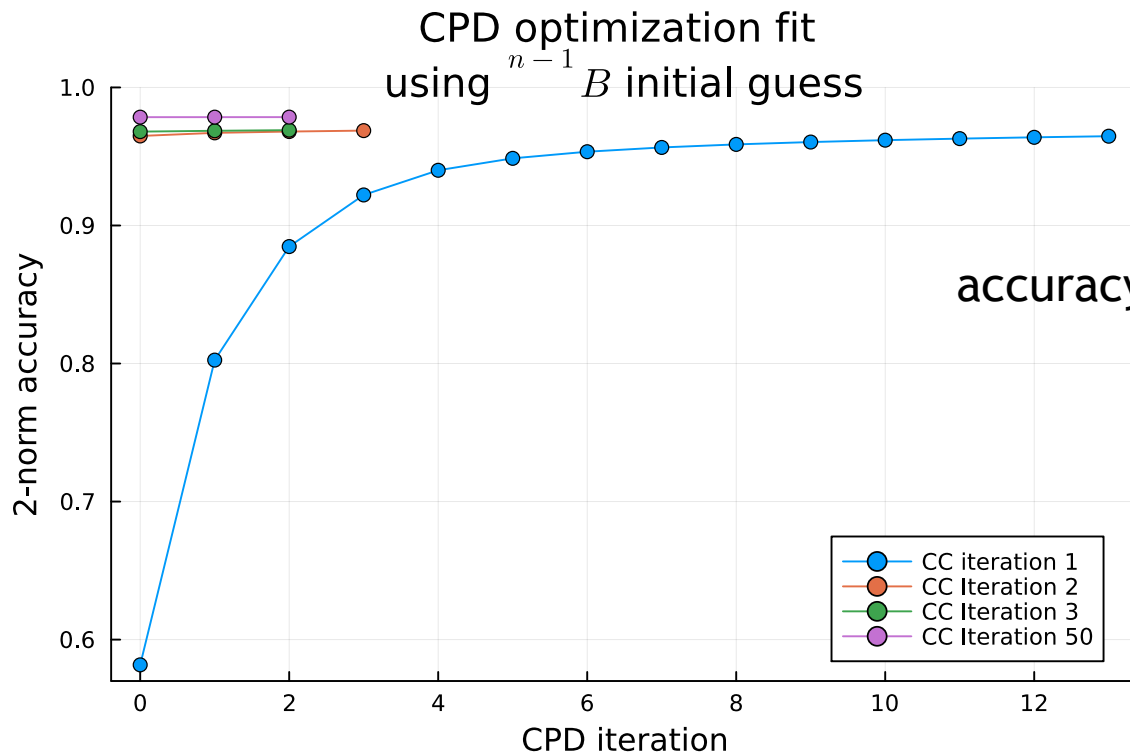
Is there a way to reduce the cost of the CPD?

$(H_2O)_2$ TIP4P geometry
DZ-F12/aug-DZ-RI



Perhaps this is border-rank optimization issue.

$(H_2O)_2$ TIP4P geometry
DZ-F12/aug-DZ-RI



Perhaps this is border-rank optimization issue.

TAMARA G. KOLDA AND BRETT W. BADER

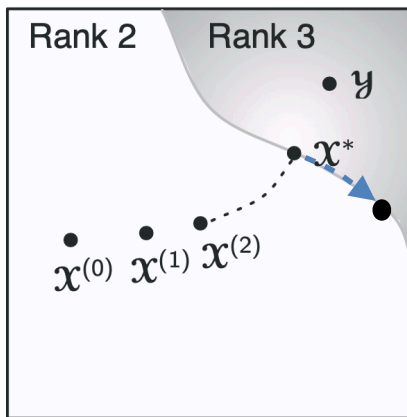
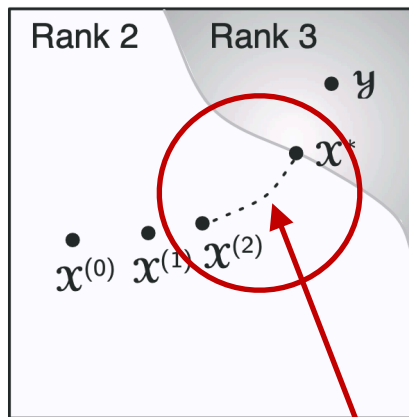


Fig. 3.2 *Illustration of a sequence of tensors converging to one of higher rank [144].*

Perhaps this is border-rank optimization issue.

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Strategy:
If CPD does 8+ iterations,
Save the factors of iteration 7
else,
Construct \hat{B} and throw factors
away



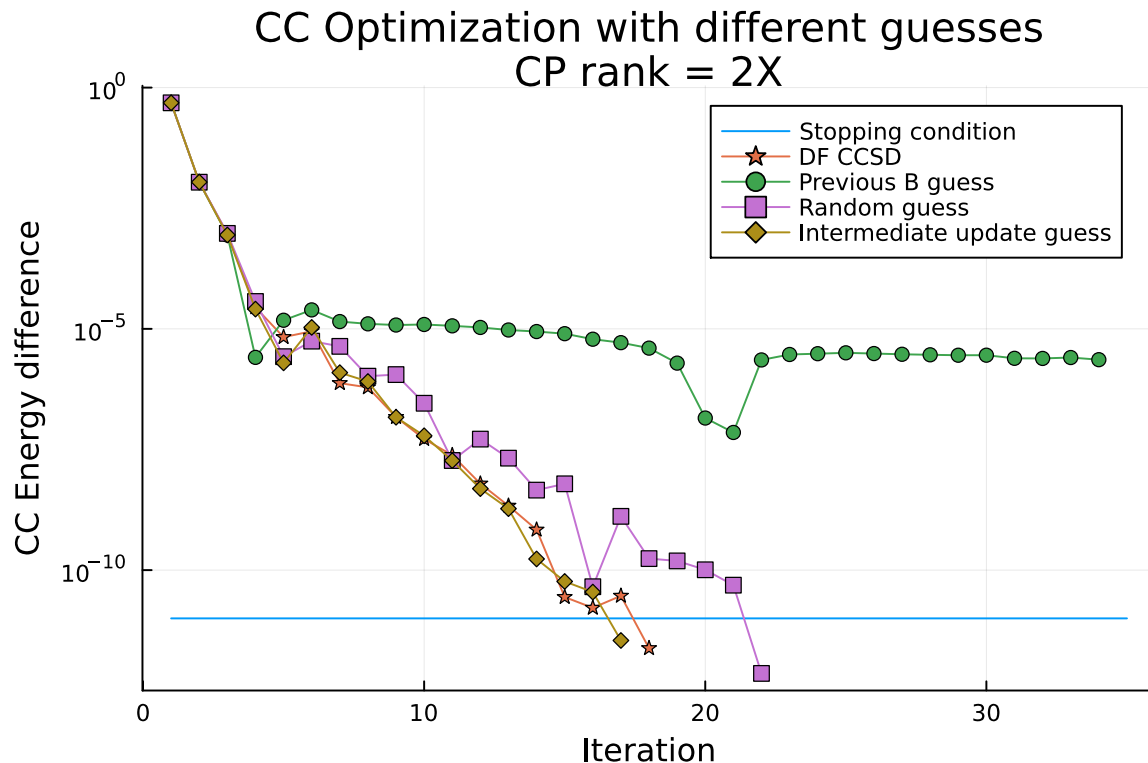
Better idea:
Take advantage of the
optimizations velocity to
put us at a “good” answer

Fig. 3.2 Illustration of a sequence of tensors converging to one of higher rank [144].

Use a set of factors from one
of these iterations

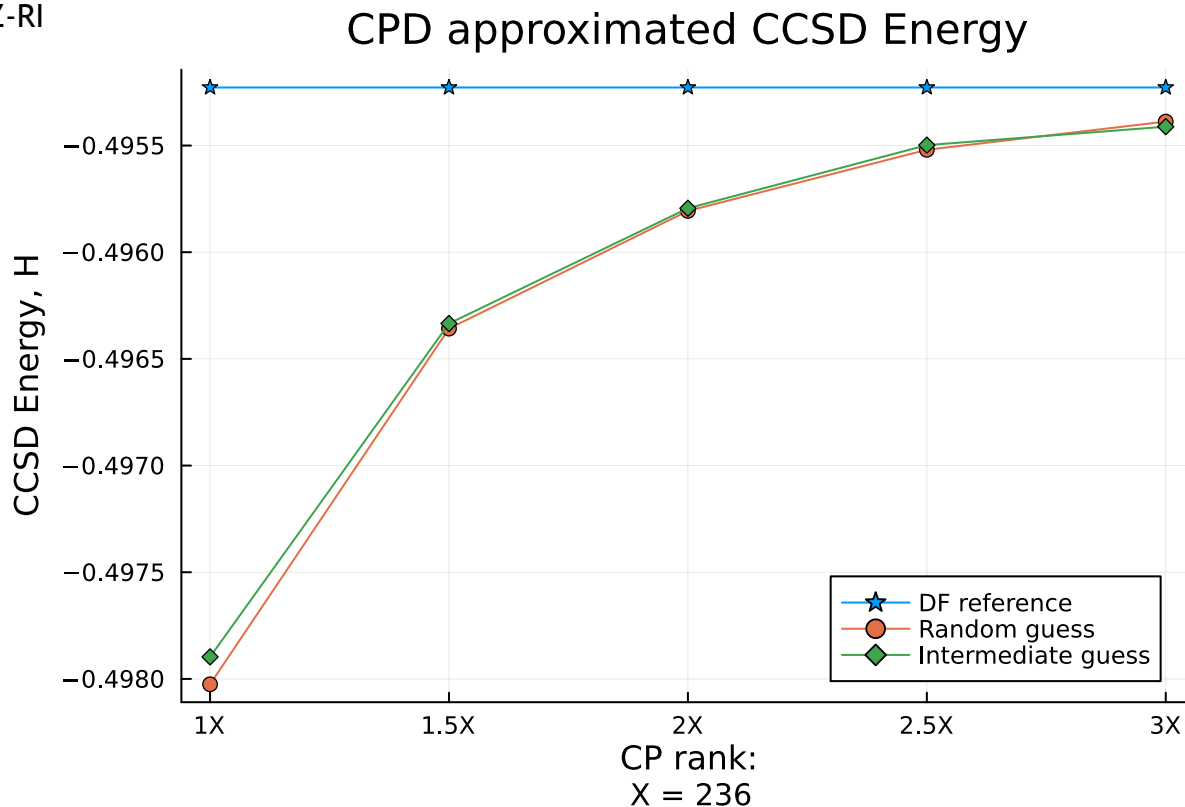
Is there a way to reduce the cost of the CPD?

$(H_2O)_2$ TIP4P geometry
DZ-F12/aug-DZ-RI



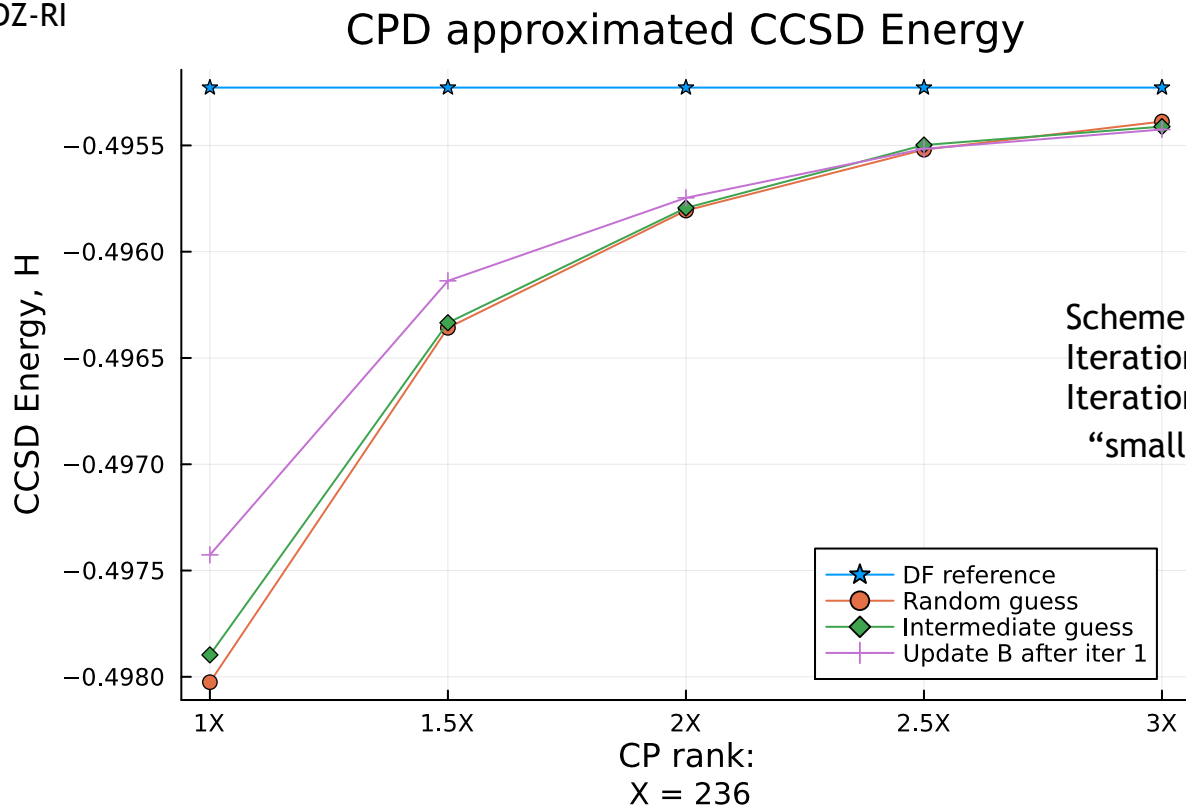
Is there a way to reduce the cost of the CPD?

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DZ-F12/aug-DZ-RI



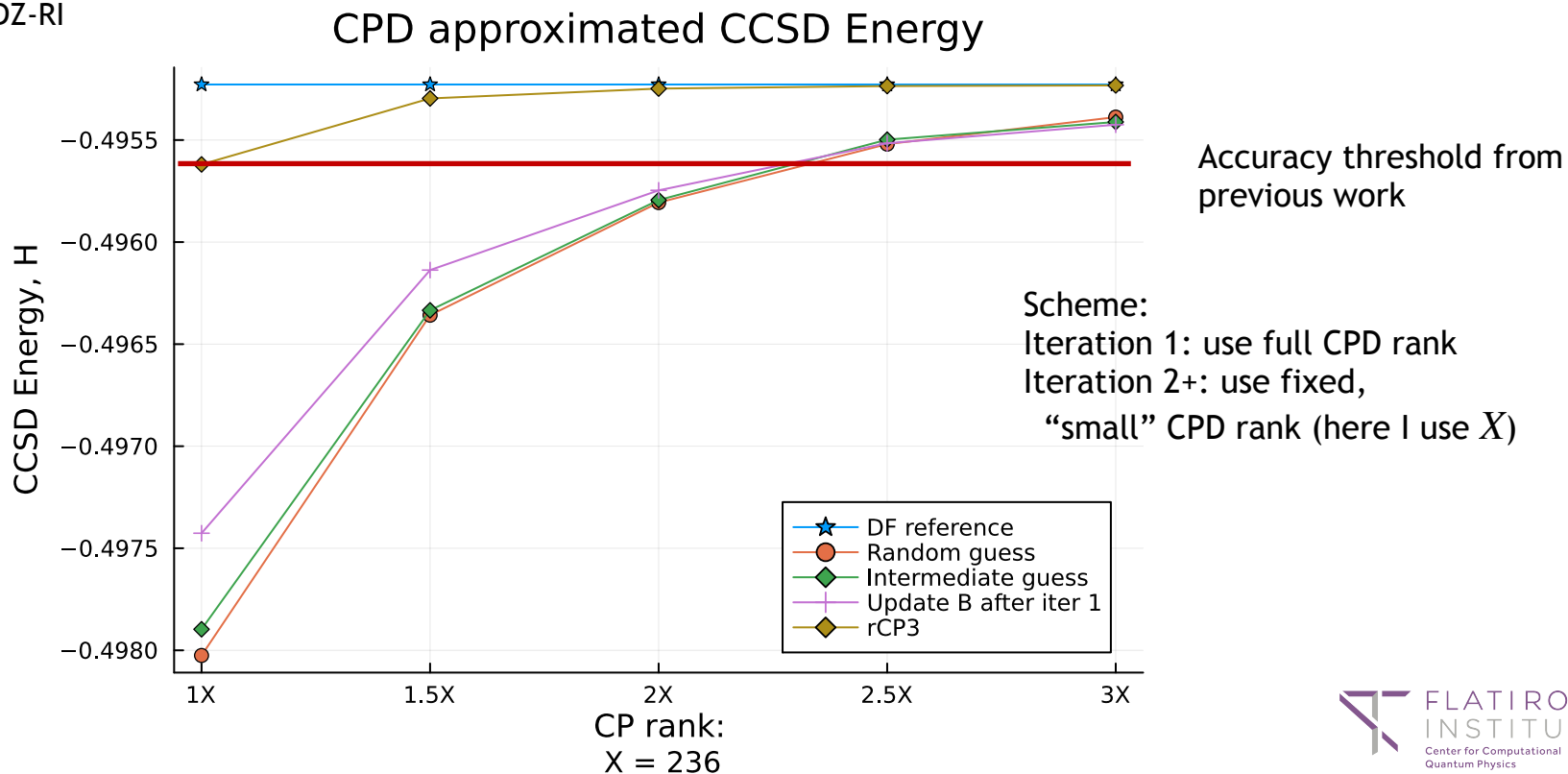
Is it necessary to recompute \hat{B} every CC iteration?

$(H_2O)_2$ TIP4P geometry
DZ-F12/aug-DZ-RI



Is it necessary to recompute \hat{B} every CC iteration?

$(H_2O)_2$ TIP4P geometry
DZ-F12/aug-DZ-RI



Summary



Mina Mandic
CCQ summer student

- It is possible to replace contractions with matrix-free tensor approximations.
- We can contrive some reliable and inexpensive way to stop matrix-free optimizations.
- Finding a way to improve the CPD accuracy for sparse tensors will directly improve accuracy of this tensor contraction.
- It could be possible to use randomized algorithms to further reduce the cost of the tensor optimization.

