

Linear-Scaling Quantum Chemistry:

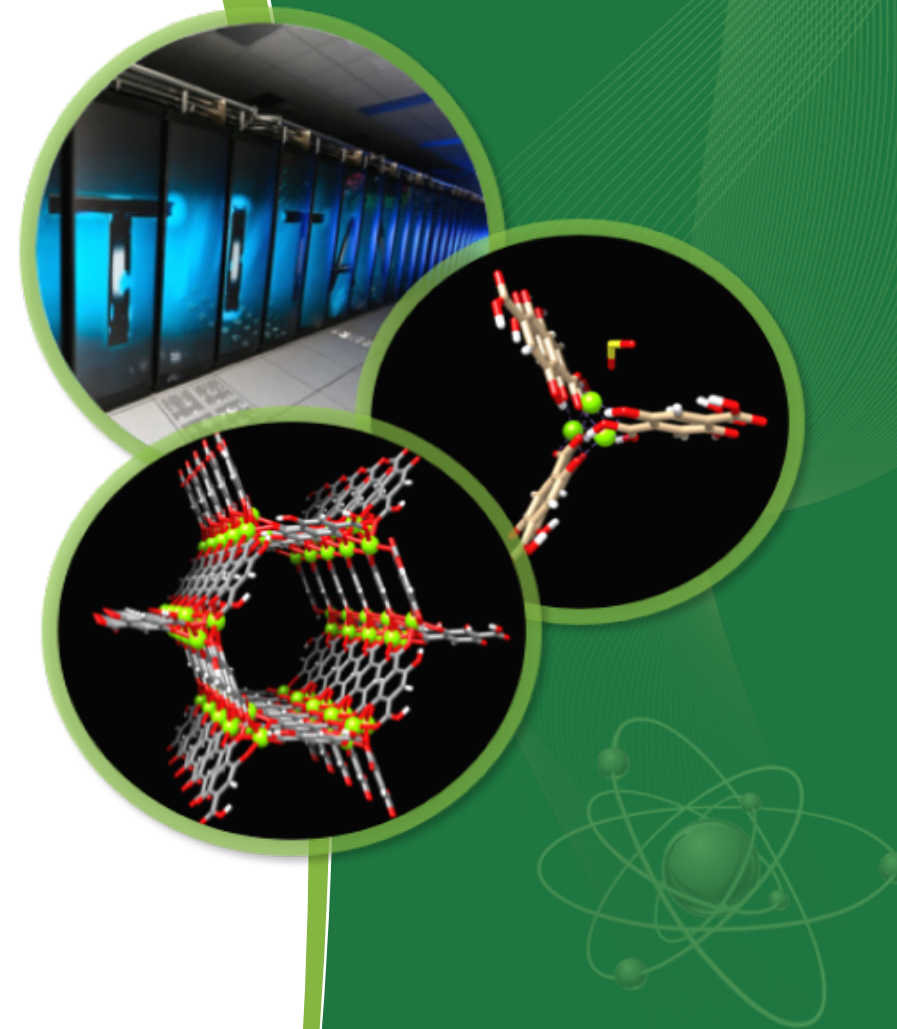
Multilayer

**Divide-Expand-Consolidate
(DEC)**

Scheme

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OLCF, ORNL

May 16, 2018



The Multilayer DEC Scheme

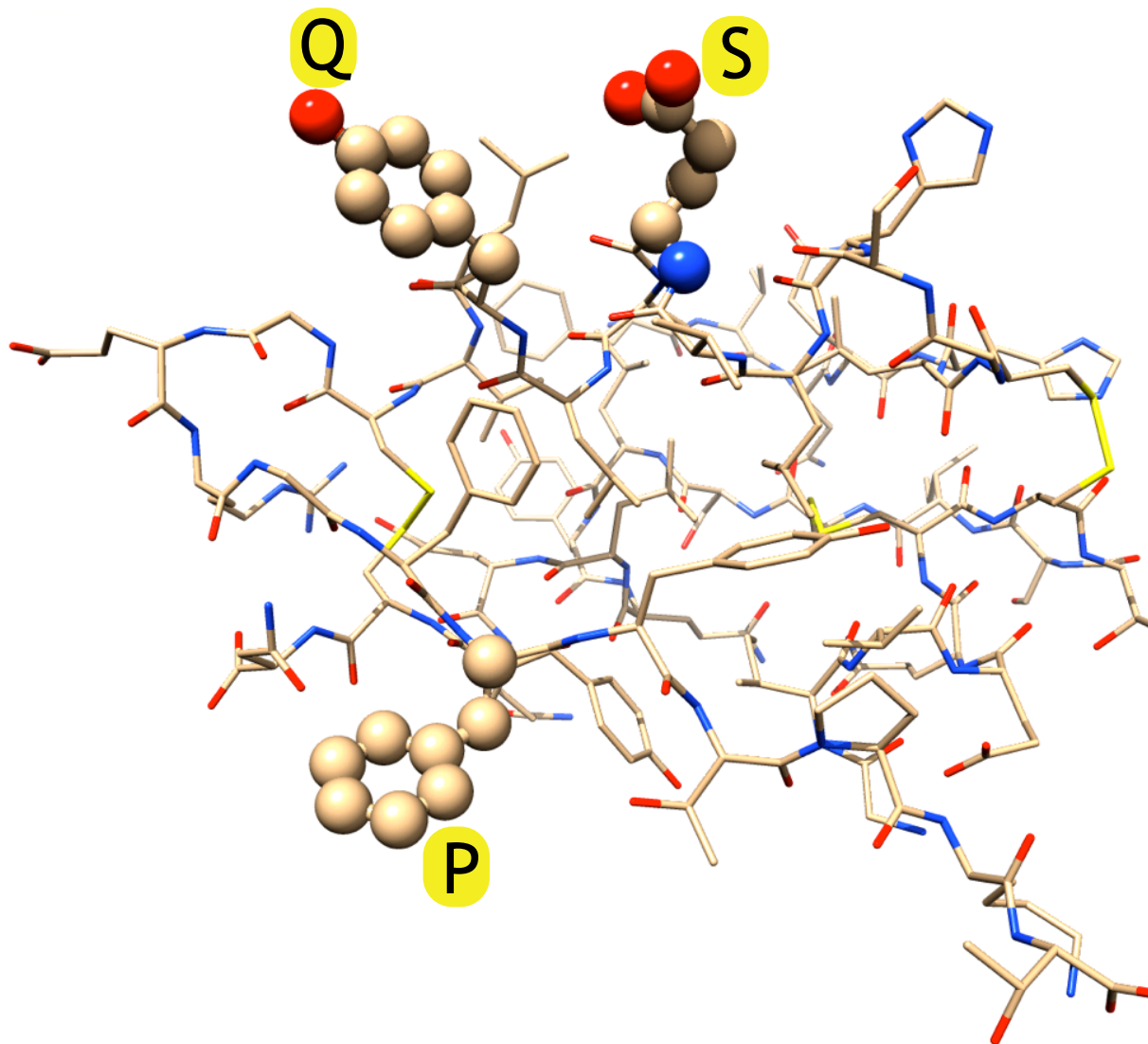


The Multilayer DEC (ML-DEC) Scheme

Small region of interest
within larger system



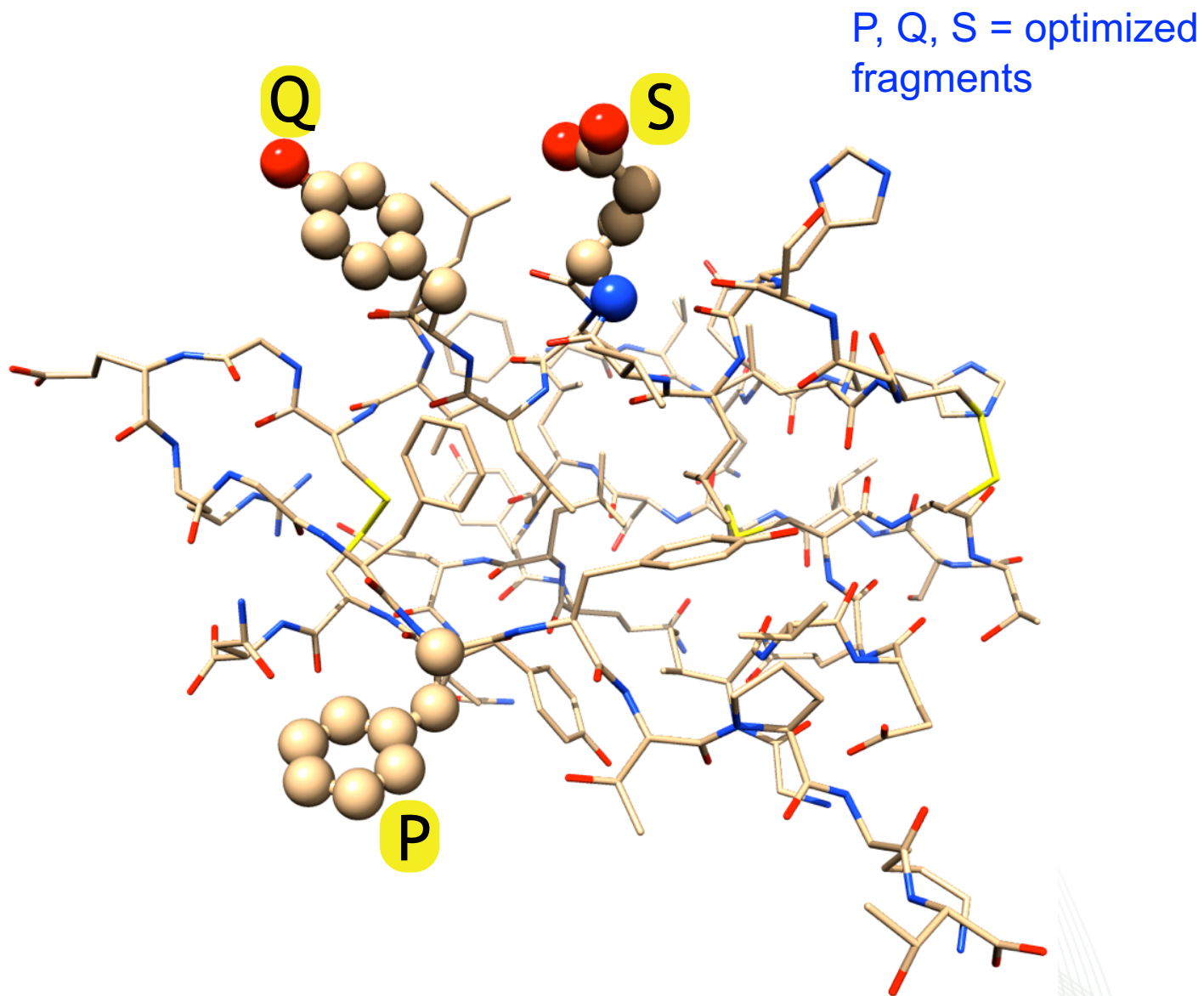
More time spent on
fragments that don't make a
strong contribution to
property of interest



Take advantage of
independent nature
of fragments

The ML-DEC Scheme

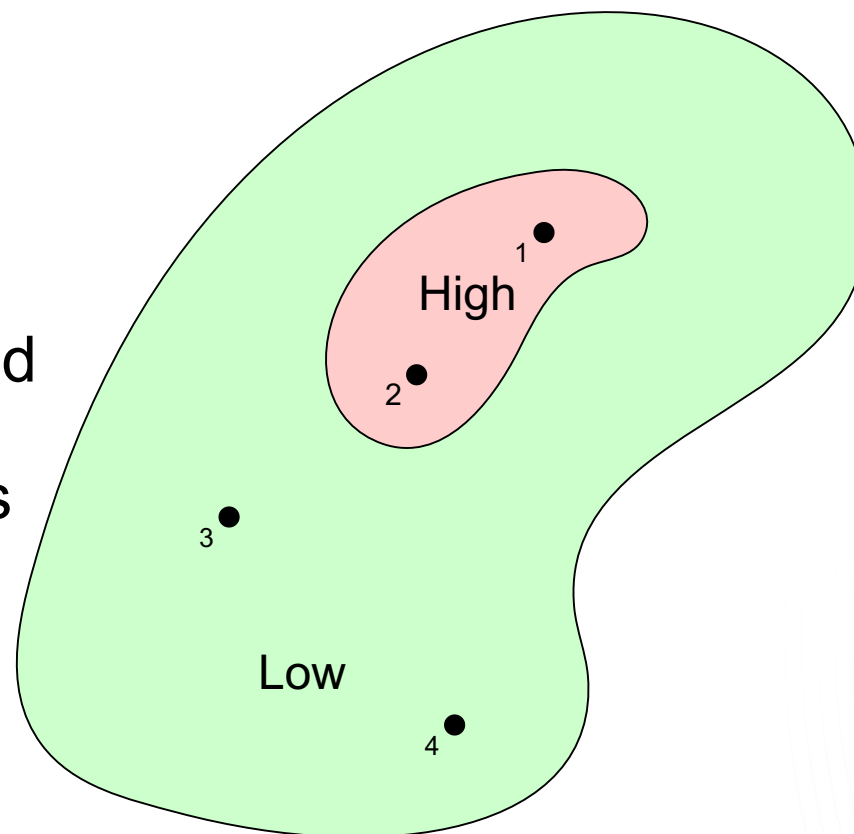
- Example: Region of interest centered on P
 - Q, S evaluated at RI-MP2 level
 - P evaluated at CCSD level
- Intent is to capture important correlation contributions in region of interest – accurate energy differences
- Suited for local effects: **adsorption, defect sites, catalytically active sites**, etc.



The ML-DEC Scheme

Low Level	High Level		
	RI-MP2	CCSD	CCSD(T)
	HF	X	
	RI-MP2		X
	CCSD		X

- User defines high and low level layers in input file and assigns atoms to each.
- All fragments optimized at RI-MP2 level of theory

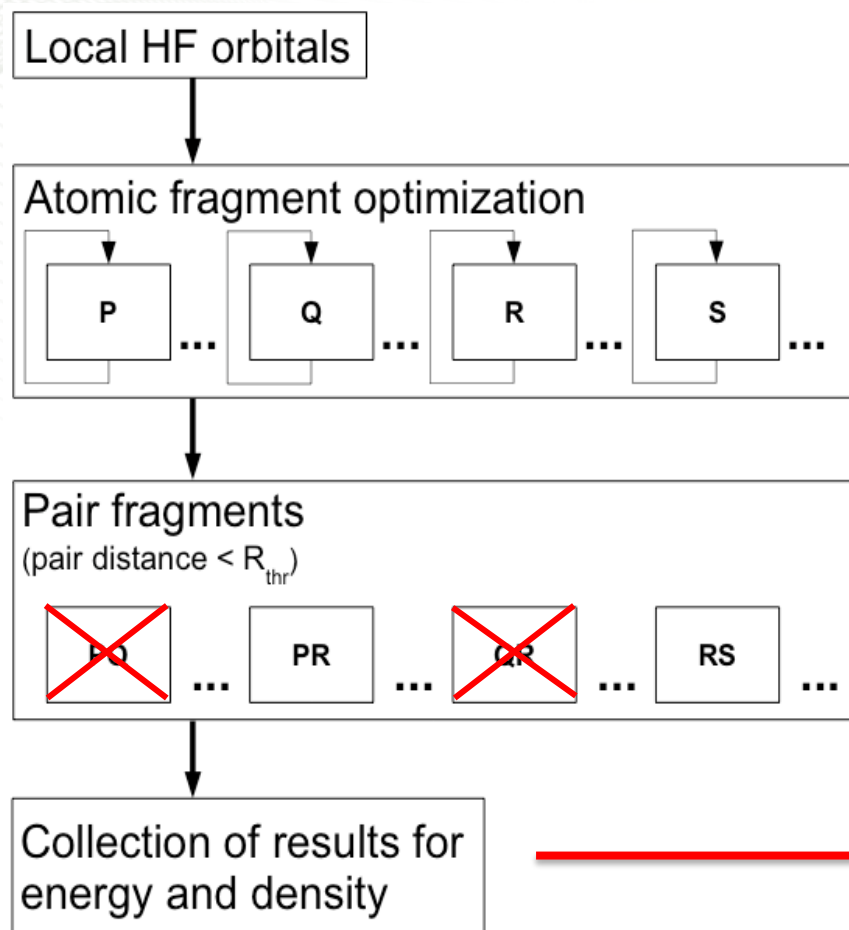


Layers not required to be contiguous
– e.g. multi-site catalysis

- After optimization, atomic fragments assigned to user-specified level of theory.
- Pair fragments are evaluated at the higher level of the two atomic fragments:
 - 1-2: CC_{high}
 - 2-4: CC_{high}
 - 3-4: CC_{low}
- Pair estimates at RI-MP2 level determine if pair level is **reduced** (not skipped).

The ML-DEC Scheme: Summary

$$E_{\text{corr}} = \sum_P^{N_{\text{frag}}} \left[E_P + \sum_{Q < P}^{N_{\text{frag}}} \Delta E_{PQ} \right]$$



Performed at RI-MP2 level for all fragments

P = Occupied orbital space

$[P]$ = Virtual orbital space

$$E_P = \sum_{\substack{i \in P, j \in P \\ ab \in [P]}} (t_{ij}^{ab} + t_i^a t_j^b) (2g_{iajb} - g_{ibja})$$

Performed at target level

$$\Delta E_{PQ} = \sum_{\substack{i \in P, j \in Q \\ ab \in [P] \cup [Q]}} (t_{ij}^{ab} + t_i^a t_j^b) (2g_{iajb} - g_{ibja}) + P \leftrightarrow Q \text{ term}$$

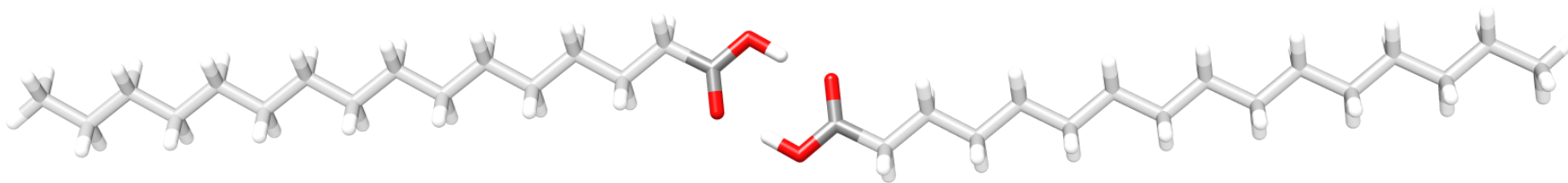
$$E_{\text{corr}} = E_{\text{corr,high}} + E_{\text{corr,low}}$$

M. Ziolkowski, B. Jansík, T. Kjærgaard, and P. Jørgensen, J. Chem. Phys. **133**, (2010).

T. Kjærgaard, P. Baudin, D. Bykov, K. Kristensen, and P. Jørgensen, Wiley Interdiscip. Rev. Comput. Mol. Sci. (2017).

Test Case: Palmitic Acid Dimer

Palmitic acid dimer
cc-pVTZ, FOT=10⁻⁴
1976 basis functions



HF/RIMP2, 10.0 Å cutoff

Model	Nproc	RI-MP2 frags	Wall time (min)	E _{corr}
DEC monomer	7	17*, 51	17.51	-3.241(3)
ML-DEC monomer	7	9*, 6	6.71	-1.534(2)
DEC dimer	33	36*, 130	10.19	-6.493(7)
ML-DEC dimer	33	20*, 32	5.51	-3.084(4)

*Atomic fragment optimization

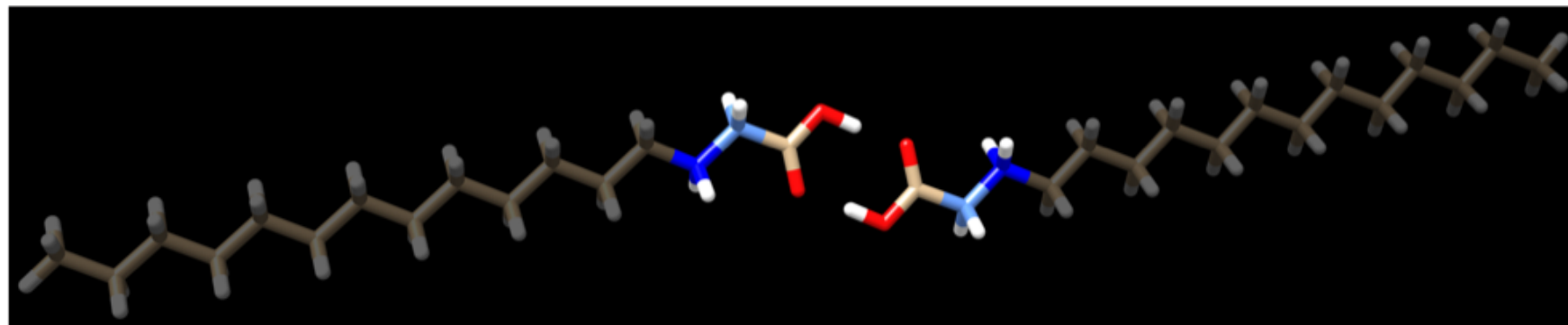
$$\Delta E_{\text{HF/RI-MP2}} = -0.033$$

$$\Delta E_{\text{RI-MP2}} = -0.027$$

Test Case: Palmitic Acid Dimer

Palmitic acid dimer
cc-pVTZ, FOT=10⁻⁴
1976 basis functions

HF/RI-MP2



Monomer, 988 basis functions

Model	Fragment Optimizations	Pair Energy Calculations
ML-DEC	9	6
ML-DEC (ext ₁)	10	9
ML-DEC (ext ₂)	11	12
DEC-RI-MP2	17	51

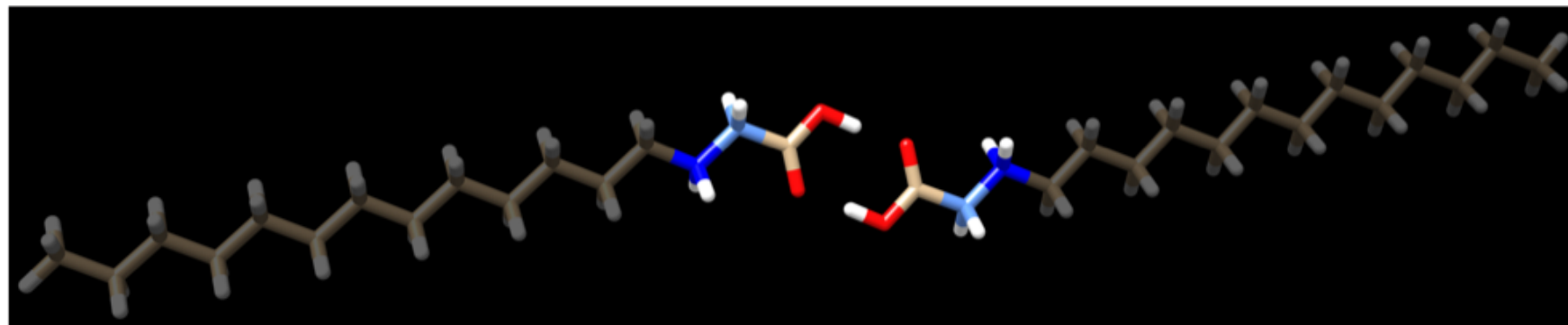
Dimer, 1976 basis functions

Model	Fragment Optimizations	Pair Energy Calculations
ML-DEC	20	32
ML-DEC (ext ₁)	22	42
ML-DEC (ext ₂)	24	50
DEC-RI-MP2	36	130

Test Case: Palmitic Acid Dimer

Palmitic acid dimer
cc-pVTZ, FOT= 10^{-4}
1976 basis functions

HF/RI-MP2

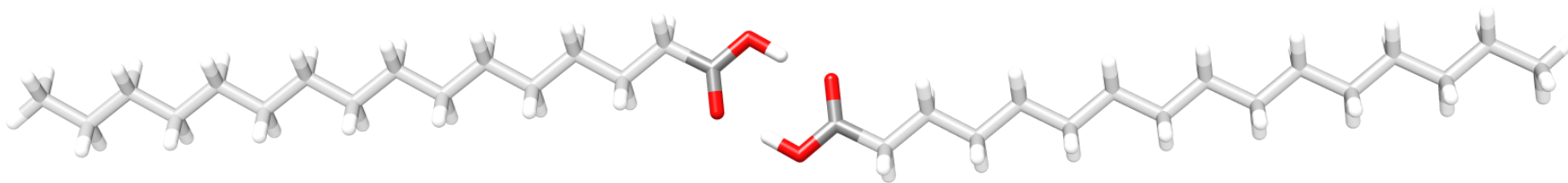


- Even with smallest high level layer, interaction energies agree within FOT.
- Agreement converges to standard DEC RI-MP2 energy as high level is increased.

Model	TTS _m (min)	TTS _d (min)	ΔE (E_h)
ML-DEC	6.71	5.51	-0.033(5)
ML-DEC (ext ₁)	7.98	6.21	-0.026(5)
ML-DEC (ext ₂)	9.27	6.84	-0.027(6)
DEC-RI-MP2	17.51	10.19	-0.027(8)

Test Case: Palmitic Acid Dimer

Palmitic acid dimer
cc-pVDZ, FOT=10⁻⁴
824 basis functions



RI-MP2/CCSD

Model	Nproc	CCSD frags	Wall time (h)	E _{corr}
DEC monomer	129	76	3.21	-2.853(3)
ML-DEC monomer	129	14	0.24	-2.655(3)
DEC dimer	129	164	8.4	-5.712(7)
ML-DEC dimer	129	38	1.85	-5.316(7)

$$\Delta E_{\text{RI-MP2/CCSD}} = -0.027$$

$$\Delta E_{\text{CCSD}} = -0.026$$

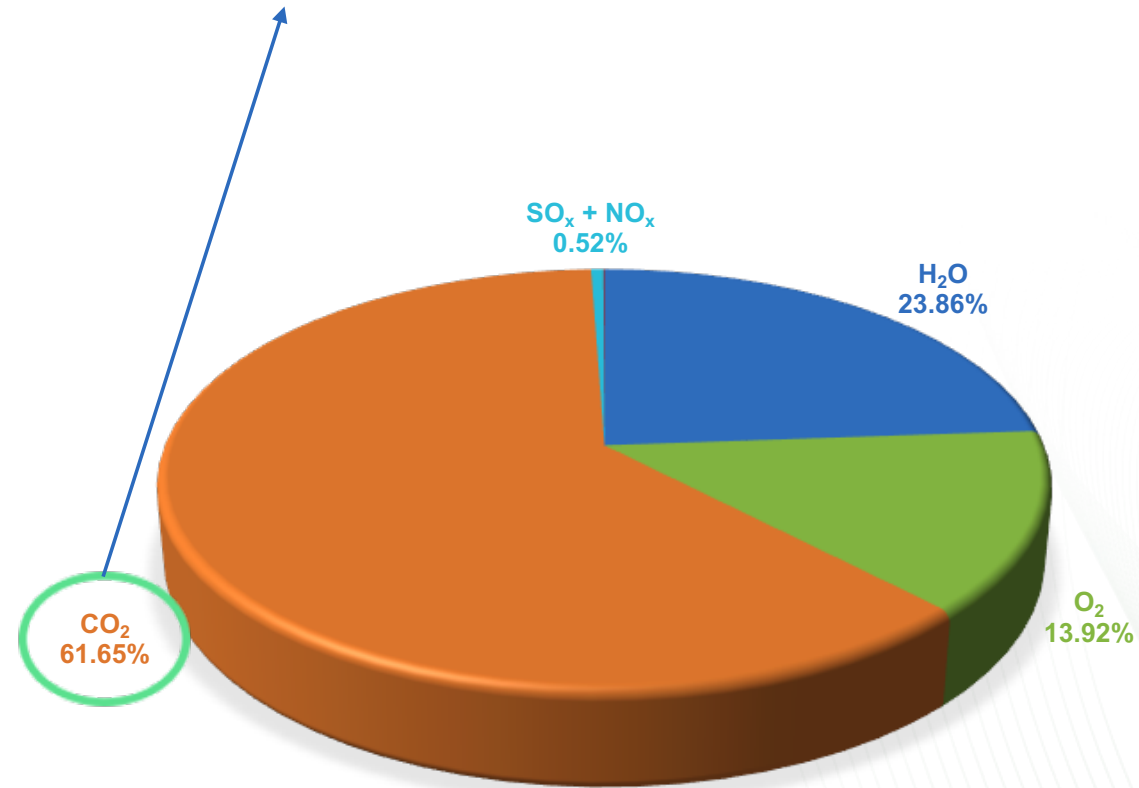
Application: Acid gas adsorption in Mg- MOF-74



MOFs for CO₂ Capture

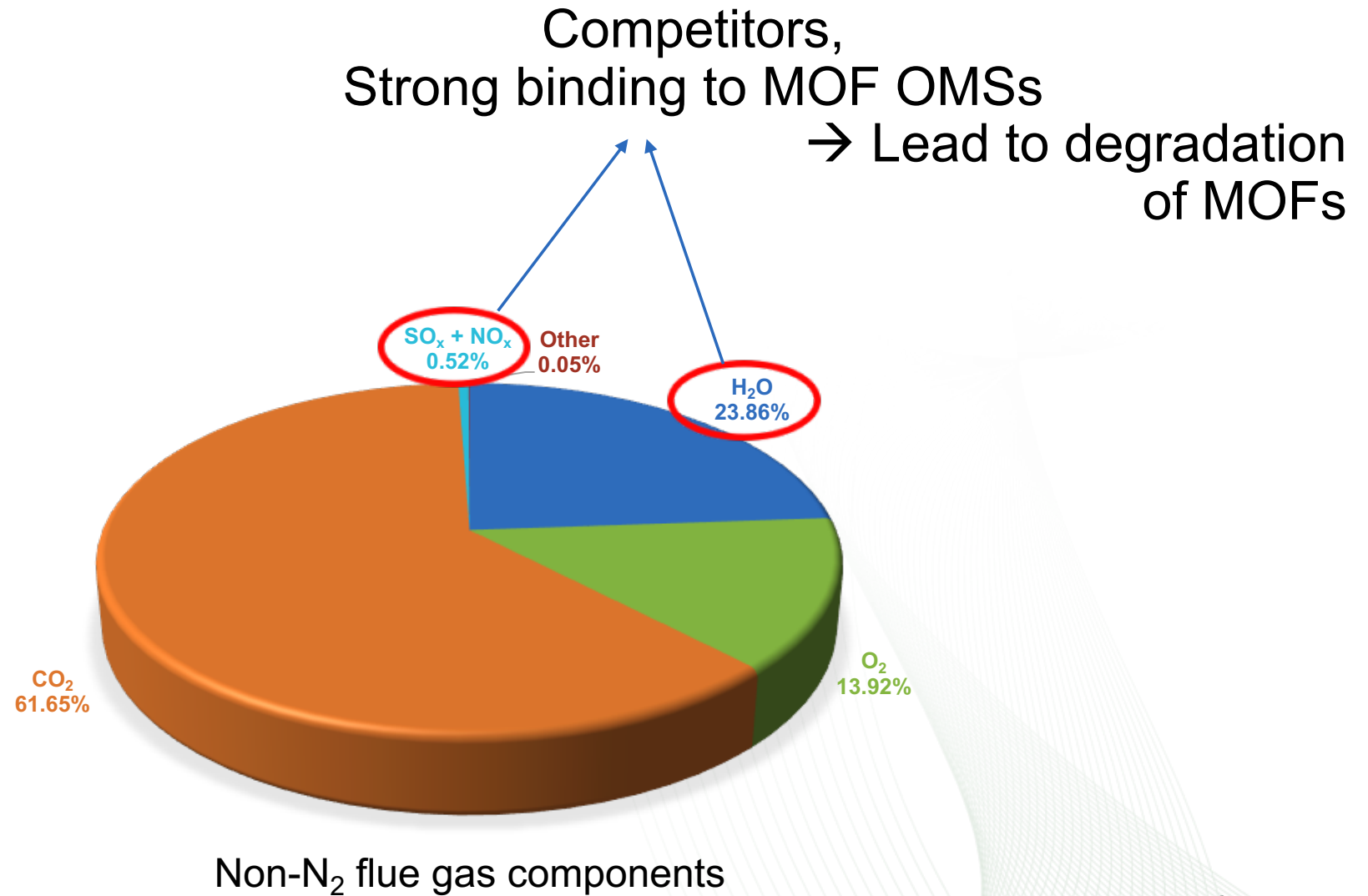


High uptake in many MOFs with OMSs
(e.g. Mg-MOF-74, HKUST-1)

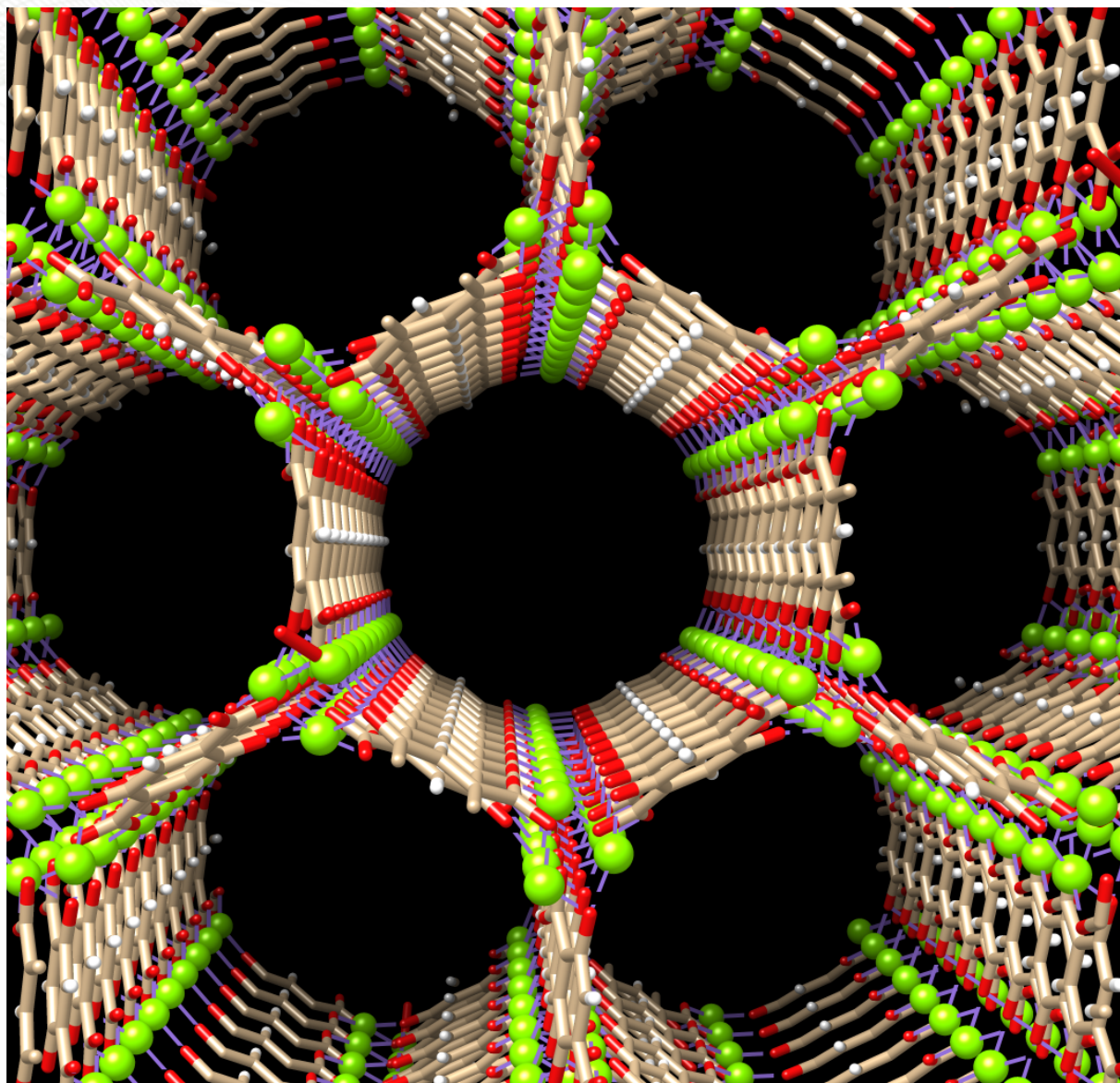


Non-N₂ flue gas components

MOFs for CO₂ Capture

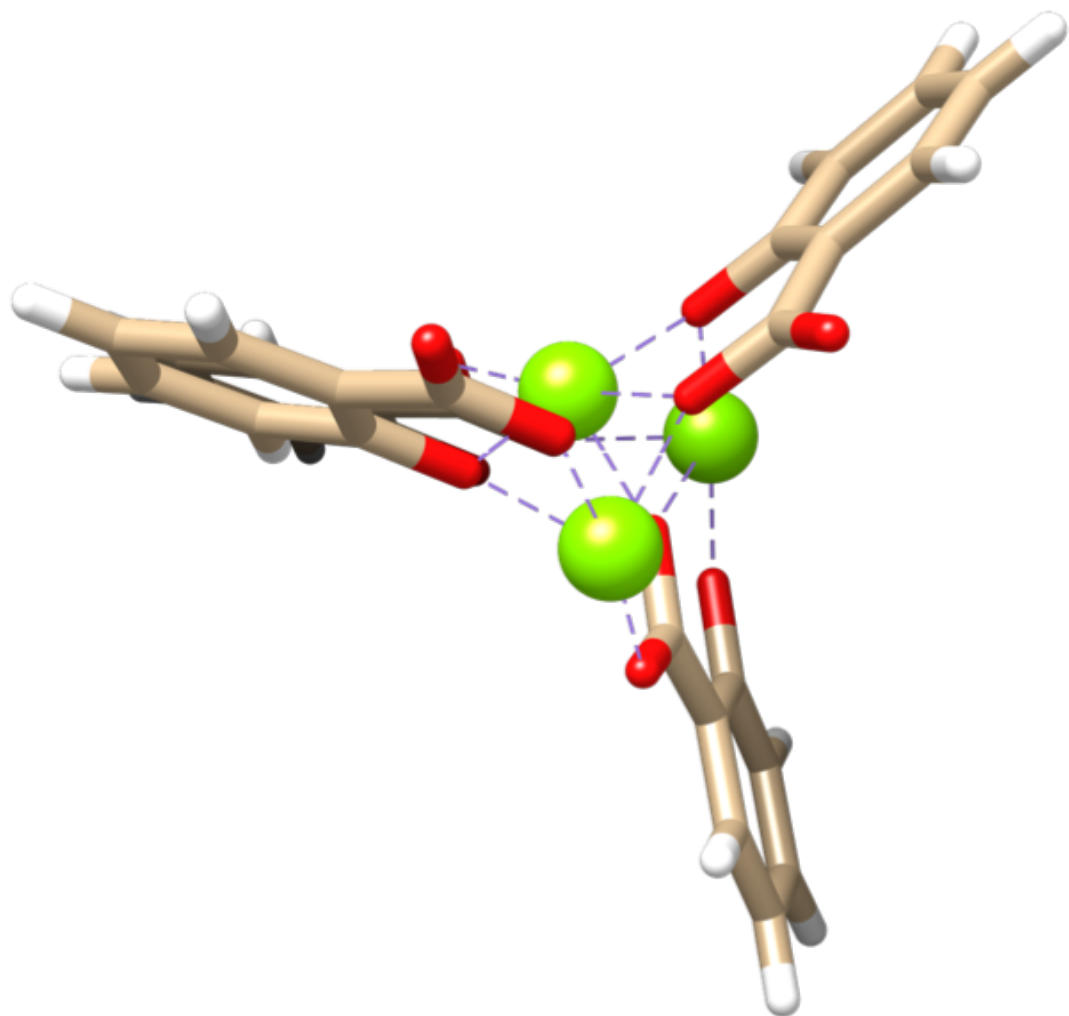


Modeling Acid Gas adsorption in Mg-MOF-74



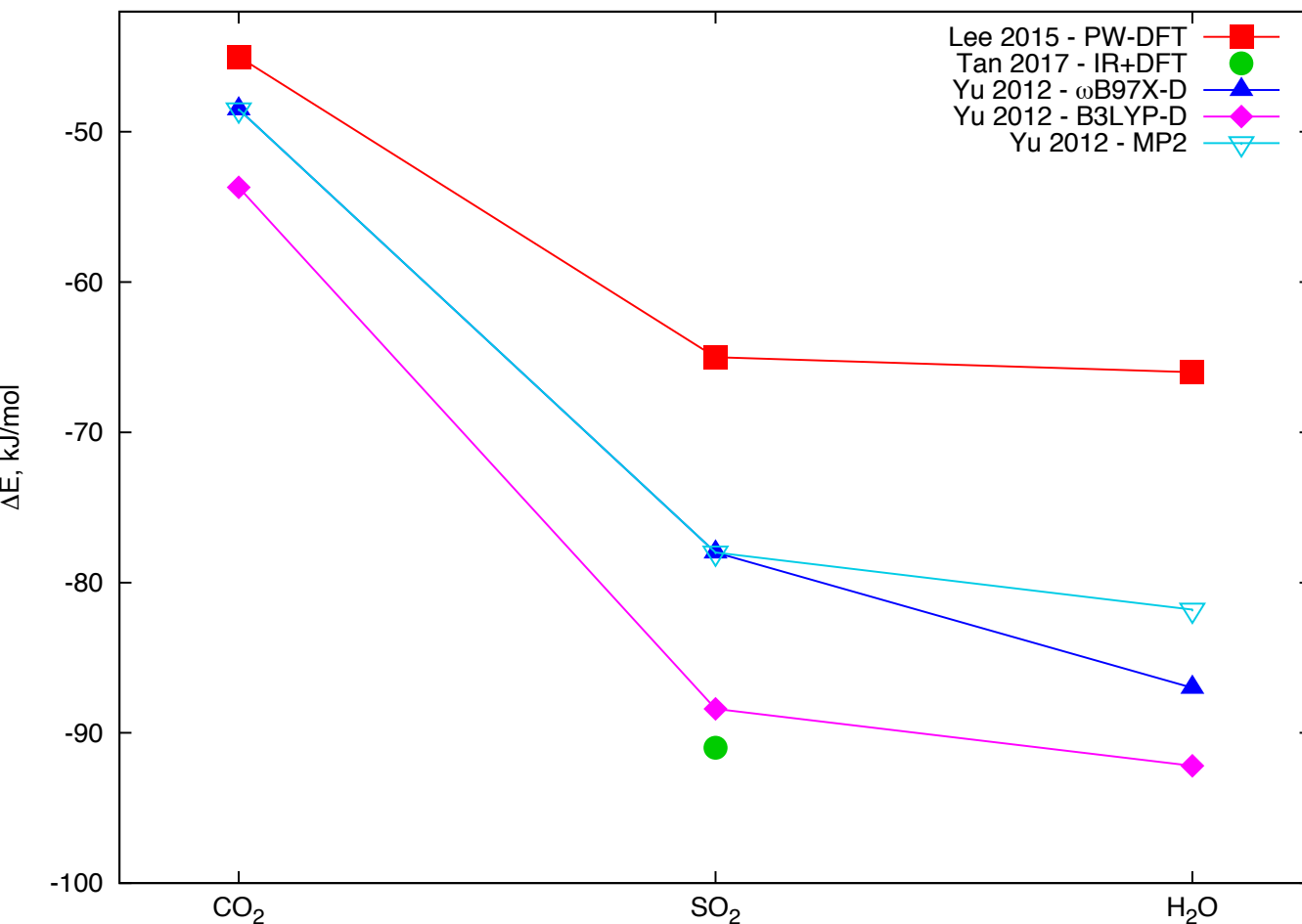
- Plane-wave DFT
 - Results depend on choice of functional, Hubbard U corrections, dispersion corrections etc.
- Small cluster models
 - Suffer from finite size effects. Clusters large enough to replicate environment within the bulk MOF are too expensive for CC.
- Desire to model large finite MOF clusters at CC level of theory while capturing bulk MOF environment.
 - Local interaction – good candidate for ML-DEC scheme.

Modeling Acid Gas adsorption in Mg-MOF-74



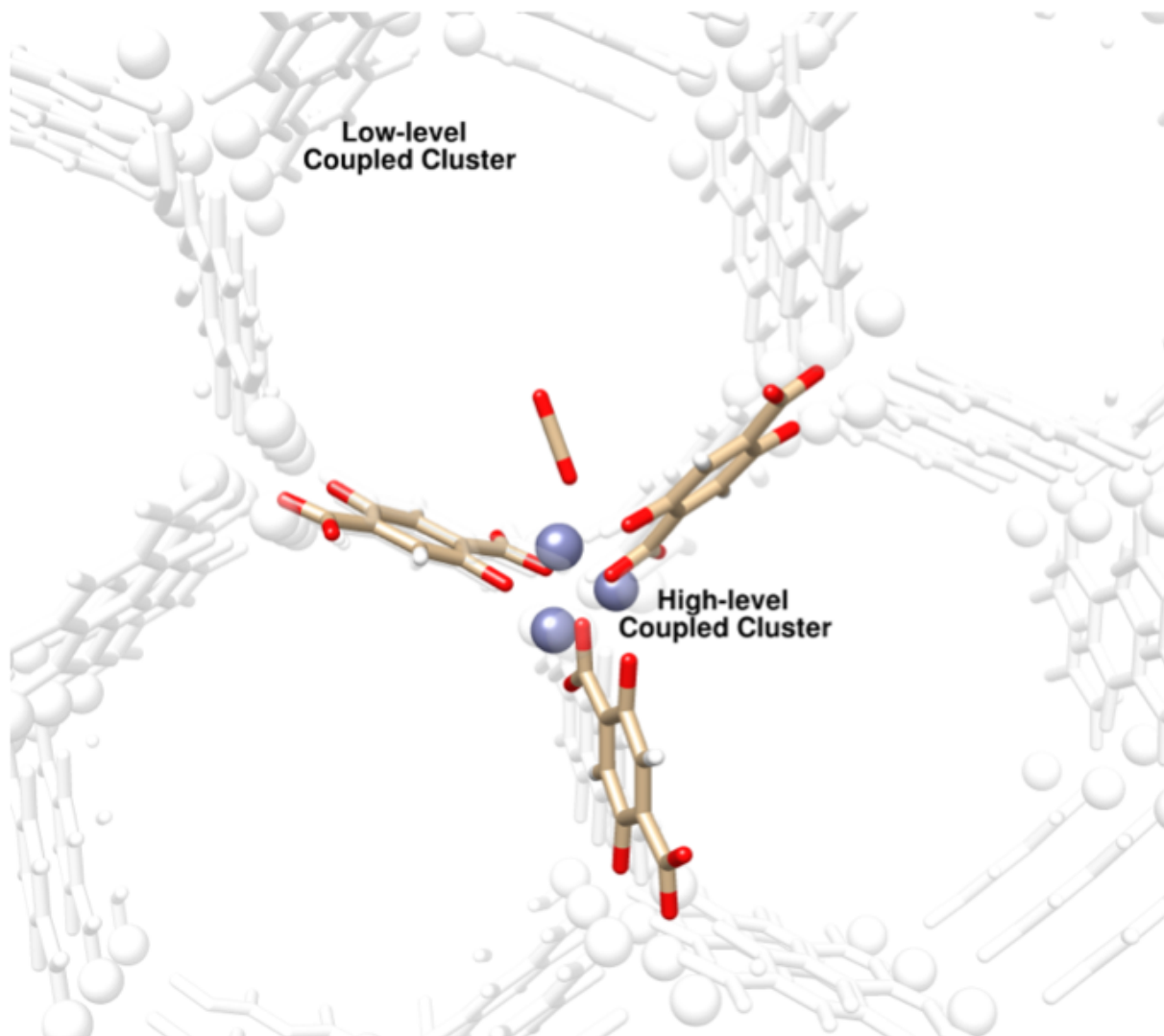
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Mg-MOF-74 + CO₂

HF/RI-MP2

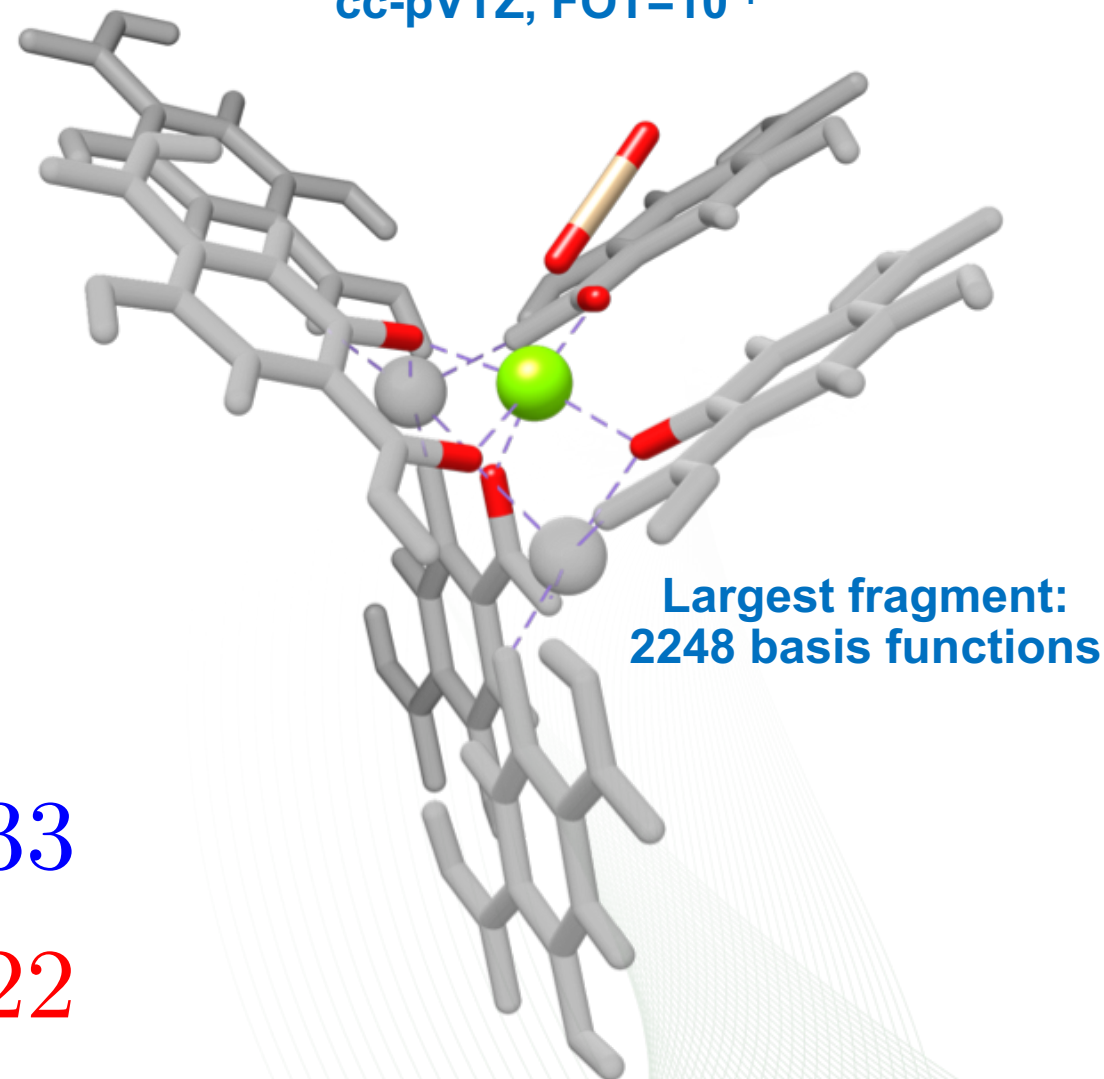
Model	NProc	RI-MP2 frags	Wall time (h)	E _{corr}
DEC MOF	65	76*, 633	3.06	-16.131(15)
ML-DEC MOF	65	76*, 135	2.26	-12.876(15)
DEC MOF+CO ₂	65	78*, 689	3.78	-16.745(16)
ML-DEC MOF+CO ₂	65	78*, 184	2.84	-13.502(16)

*Atomic fragment optimization

$$\Delta E_{\text{HF/RI-MP2}} = -0.033$$

$$\Delta E_{\text{RI-MP2}} = -0.022$$

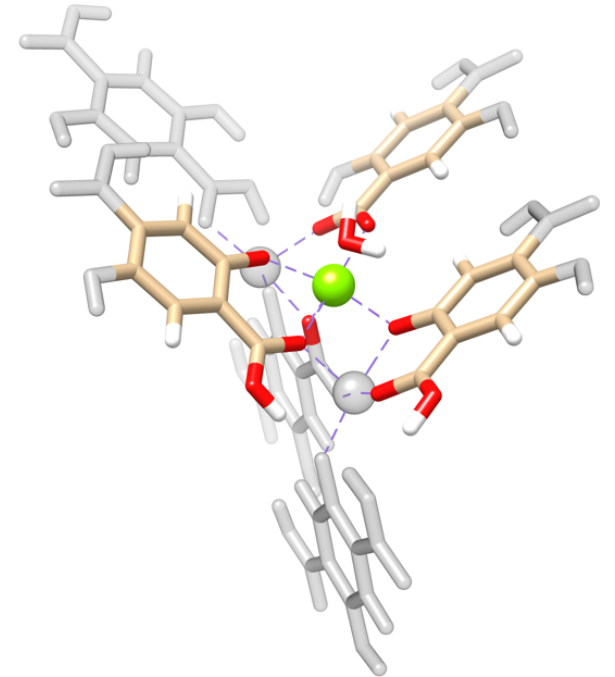
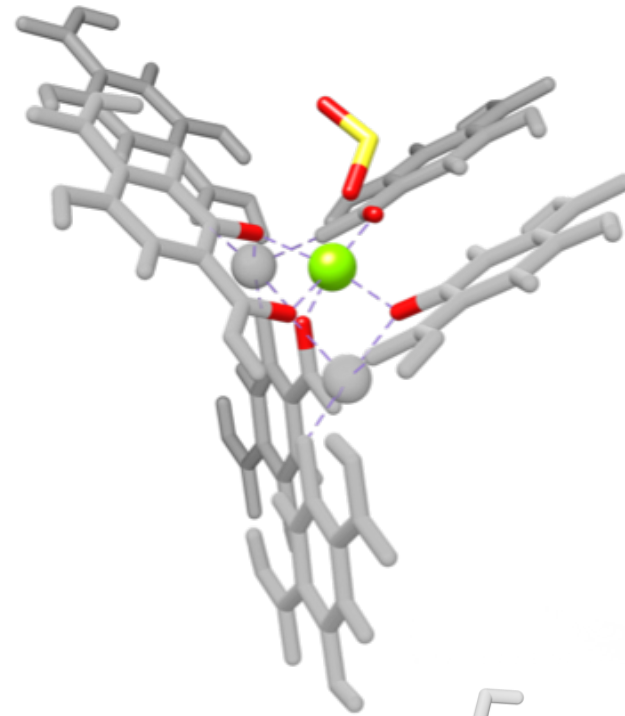
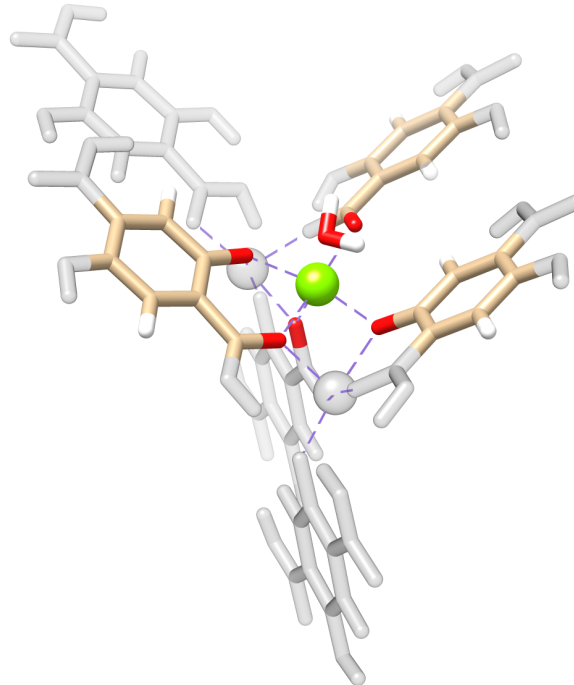
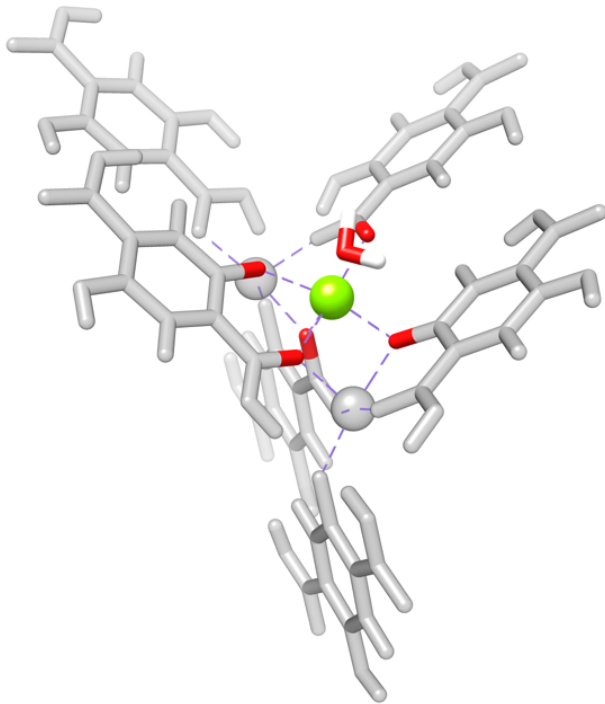
3132 basis functions
cc-pVTZ, FOT=10⁻⁴



Mg-MOF-74 + SO₂/H₂O

Interaction energy (E_n)

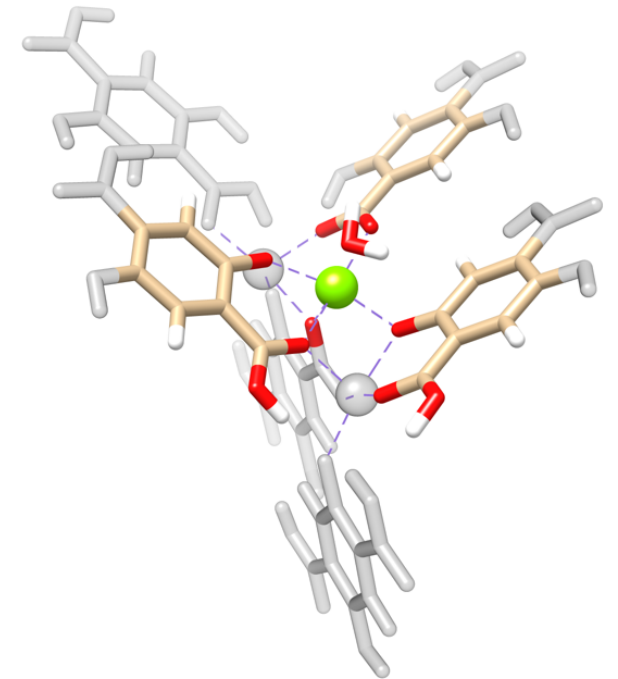
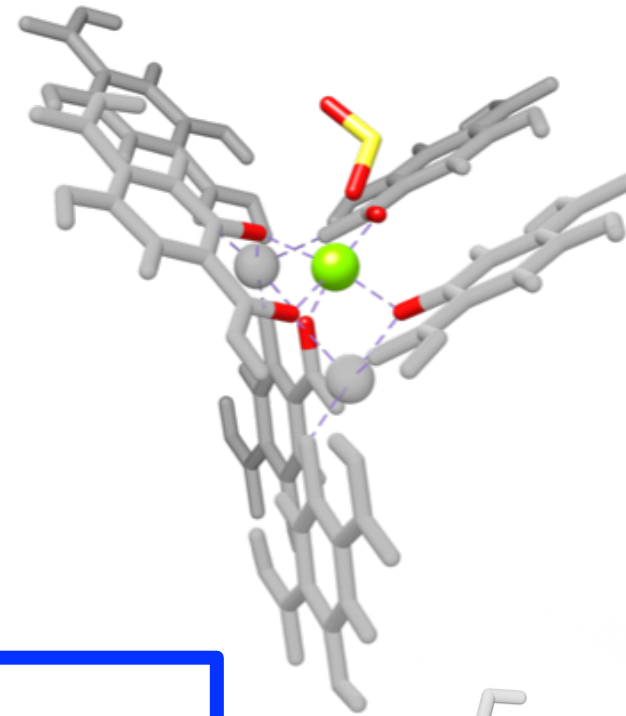
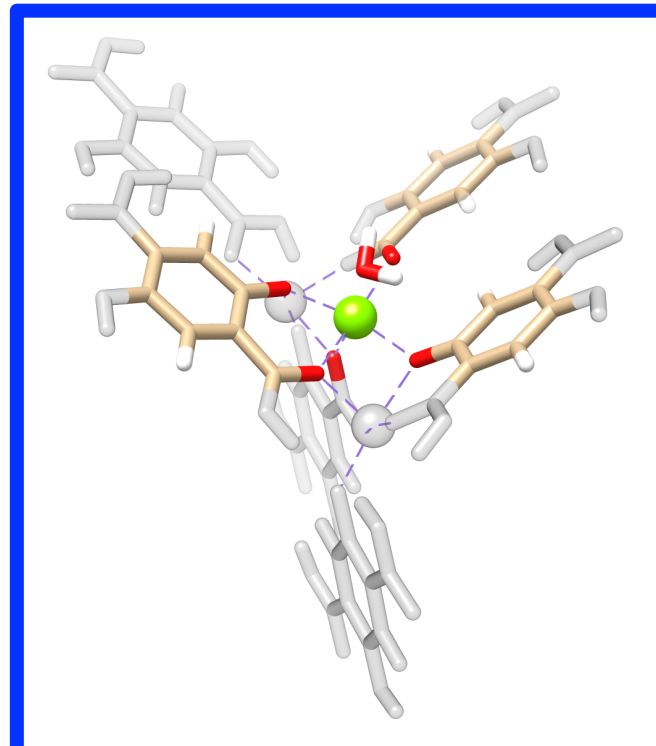
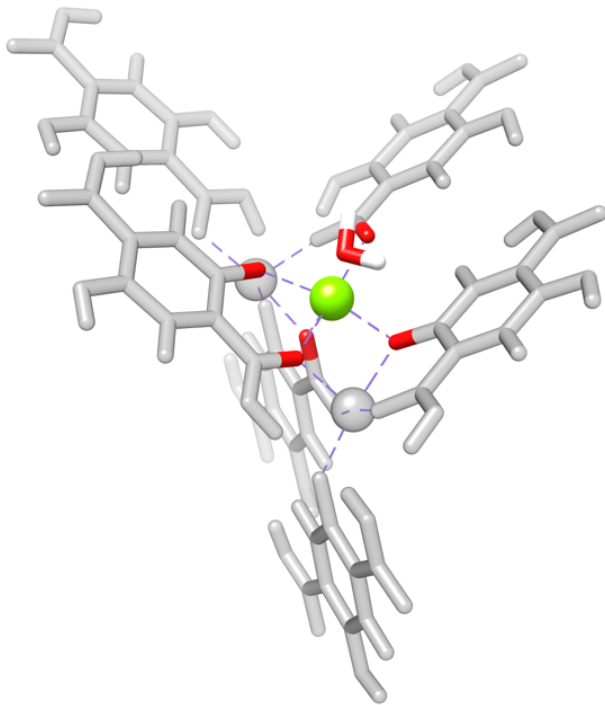
Gas	HF/RI-MP2	RI-MP2
SO ₂	0.028(16)	-0.029(16)
H ₂ O	-0.099(16)	-0.038(16)
CO ₂	-0.033(16)	-0.022(16)



Mg-MOF-74 + SO₂/H₂O

Interaction energy (E_n)

Gas	HF/RI-MP2	RI-MP2
SO ₂	-0.028(16)	-0.029(16)
H ₂ O	-0.073(16)	-0.038(16)
CO ₂	-0.033(16)	-0.022(16)



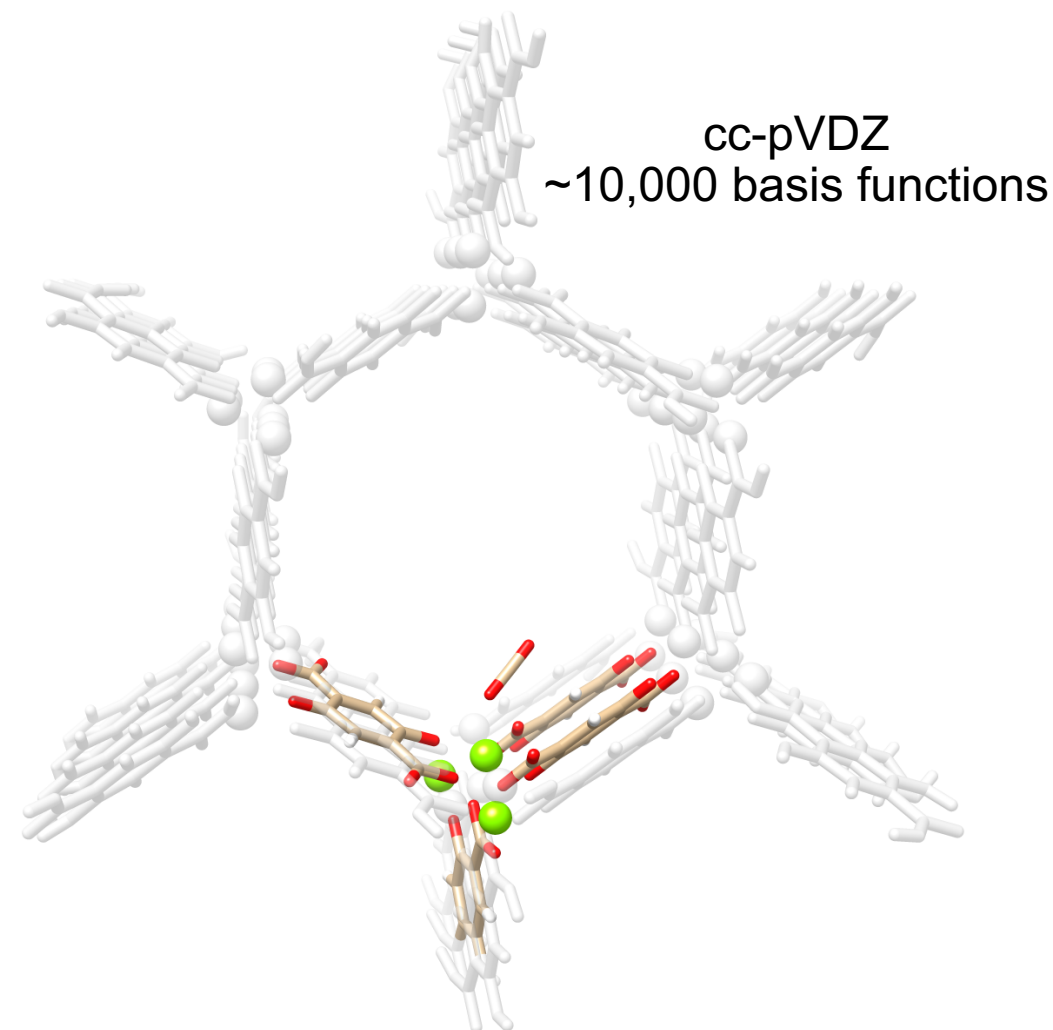
Target system: MOF channel

- HF/RI-MP2 to optimize high level layer.
- RI-MP2/CCSD for final interaction energy calculations.

Model	Fragment Optimizations	Pair Energy Calculations
ML-DEC	202	201
ML-DEC (ext ₁)	484	510
DEC-RI-MP2	568	7002

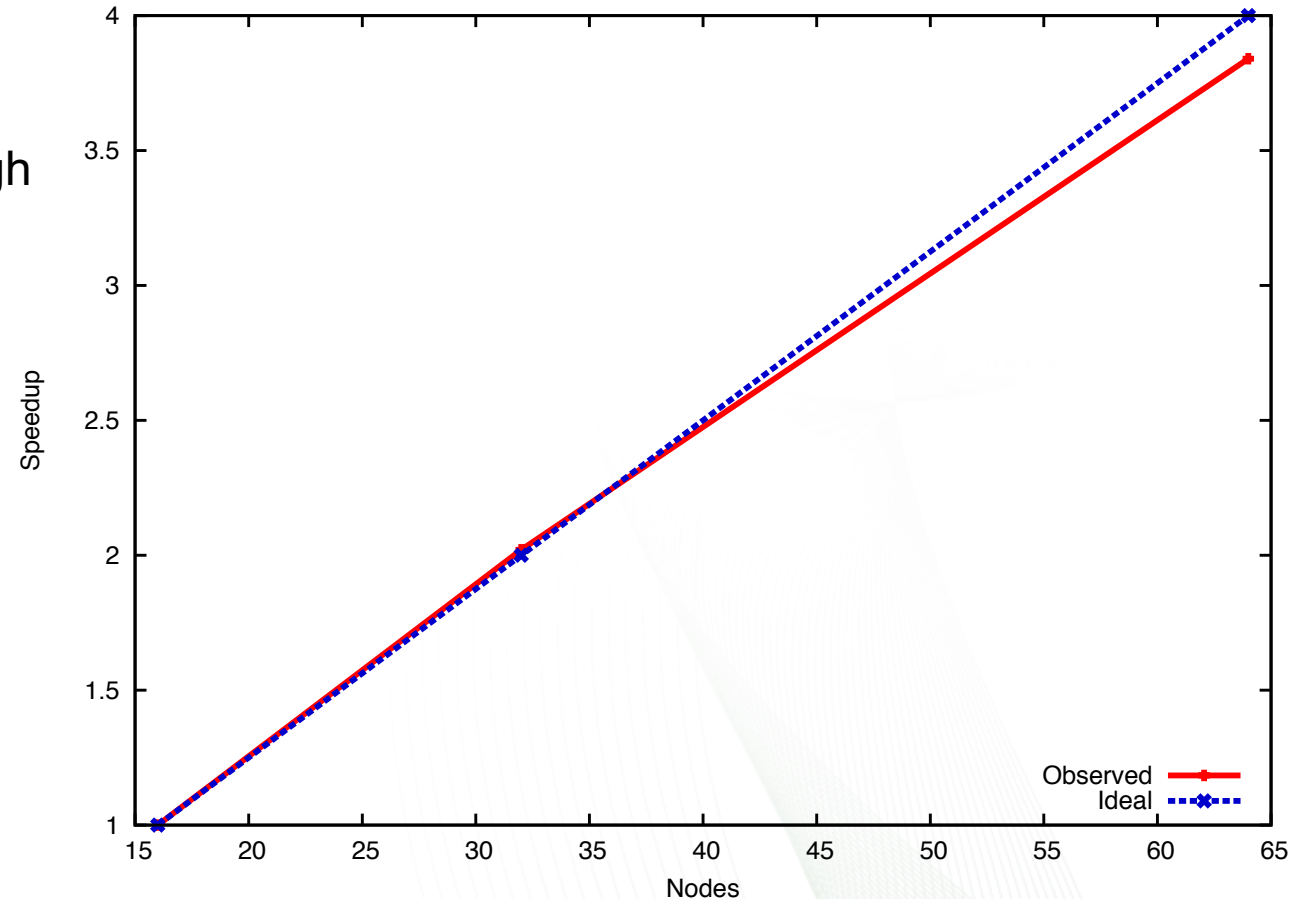
Most
important for
HF/RI-MP2

Most
important for
RI-MP2/CCSD



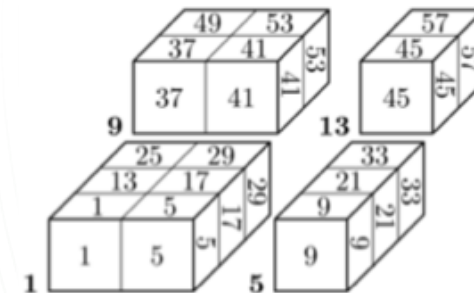
Targeting Performance on Summit

- LSDalton makes use of MPI, OpenMP, and OpenACC.
 - Linear scaling as long as problem is big enough
- RI-MP2 modules already include OpenACC acceleration
 - `jsrun -n 6 -a 1 -g 1 -r 6 -c 7 -brs`
 - >2x speedup using 1 GPU/rs compared to 0 GPUs/rs
- Current efforts focus on integration with the TAL-SH tensor algebra library for GPU-acceleration of CC modules (https://github.com/DmitryLyakh/TAL_SH)



GPU Offloading of Tensor Contractions with TAL-SH

- CC calculations require evaluation of tensor contractions. Efficient handling of these contractions is key to good performance of these modules.
- Currently utilizing [ScaTeLib: a Scalable Tensor Library](#) in order to distribute tensor contractions across multiple ranks via tiling.
- E.g.:
$$C_{ijkl} = \alpha \sum_{a,b} A_{ajlb} B_{kbai} + \beta C_{ijkl}$$
 - A and B must be sorted appropriately: $A_{ajlb} \Rightarrow A_{jlab}$, $B_{kbai} \Rightarrow B_{abki}$
 - Call dgemm
 - Returns contribution to local C tile as C_{jlki} . Reorder $\Rightarrow C_{ijkl}$ and add to local tile.



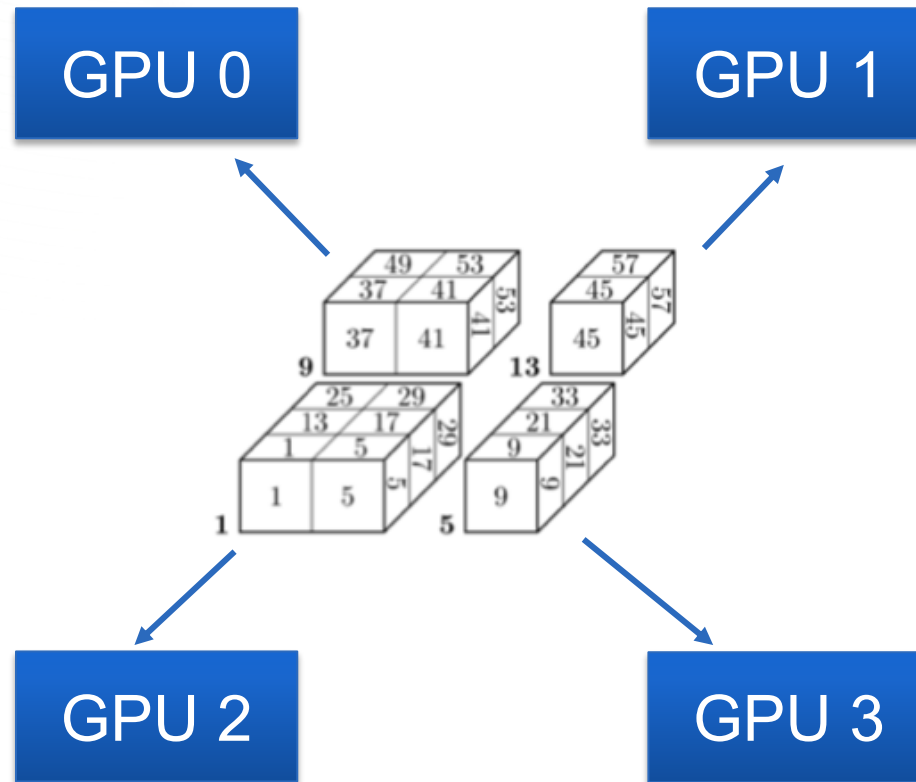
GPU Offloading of Tensor Contractions with TAL-SH

- **TAL-SH**: **T**ensor **A**lgebra **L**ibrary for **S**hared-memory systems. Integrated as backend for ScaTeLib.

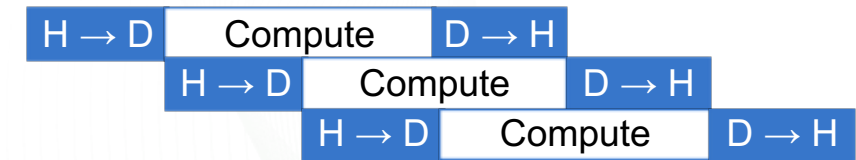


GPU Offloading of Tensor Contractions with TAL-SH

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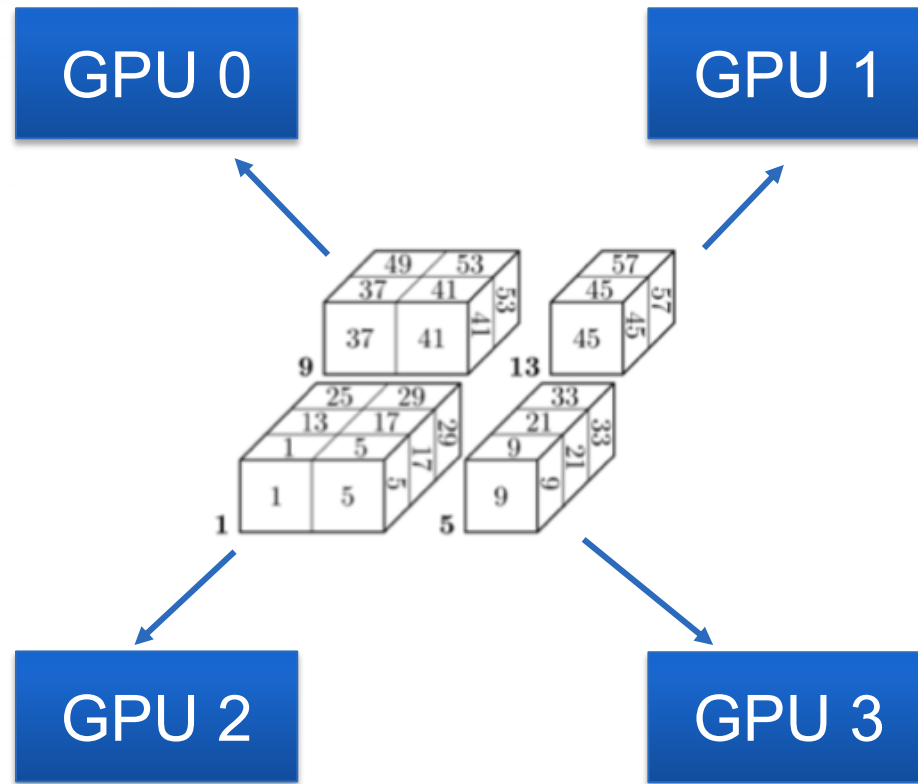
- Handles sorting, no explicit reorder calls necessary in ScaTeLib
- Asynchronous task scheduling
- Tasks are pipelined to overlap computation and data transfer
 - 2 active tasks per GPU at any time



- Basic implementation provides ~2.25x speedup for contraction of two 4th-order tensors with tile size of 80 (Titan).

GPU Offloading of Tensor Contractions with TAL-SH

- TAL-SH: Tensor Algebra Library for Shared-memory systems. Integrated as backend for ScaTeLib.



- Targeted performance improvements:
 - OpenMP compatibility
 - Reduction of data transfer
 - Optimization of tile sizes used within LSDalton
 - Ensure time spent in data transfer < time spent in computation

Acknowledgements

Dmytro Bykov: LSDALTON CAAR Liason

Dmitry Lyakh: TAL-SH (https://github.com/DmitryLyakh/TAL_SH)



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