

Investigation of the Effect of Framework Flexibility on Adsorption in SIFSIX-3-Cu Using a Machine-Learned Force Field

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Characteristics of a Good Physisorbent

- Characteristics of an economically viable sorbent

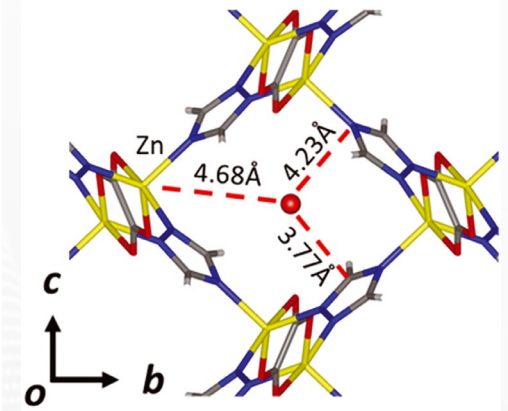
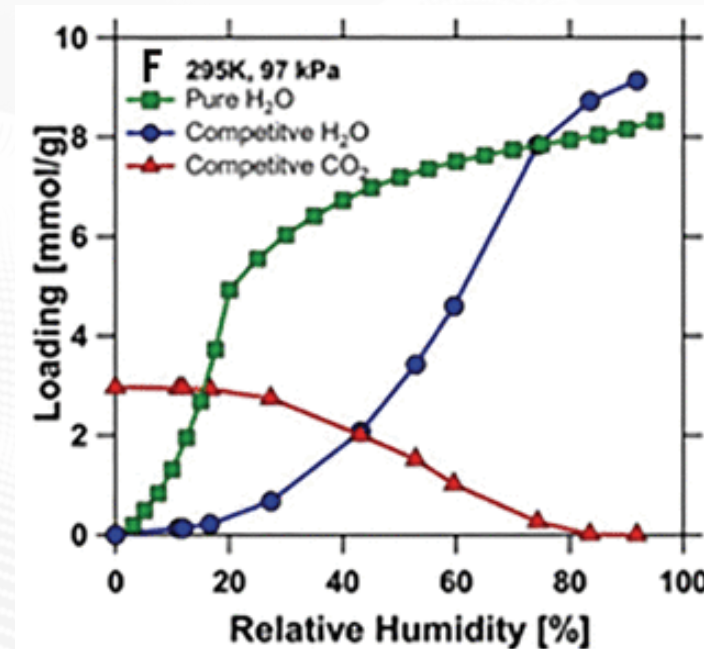
- Selective: more efficient gas separations
- Stable: lower material costs
- Regenerable: reduced cost of power consumption

- Metal-organic frameworks (MOFs)

- Metal nodes connected by organic linkers
- Highly tunable for adsorption

- Example: CALF-20

- Successfully demonstrated for enhanced resource utilization in industrial processes (e.g., cement plants) and optimizing emissions from traditional power generation assets
- Little effect of humidity up to 30% RH
 - Strong **dispersion** interactions

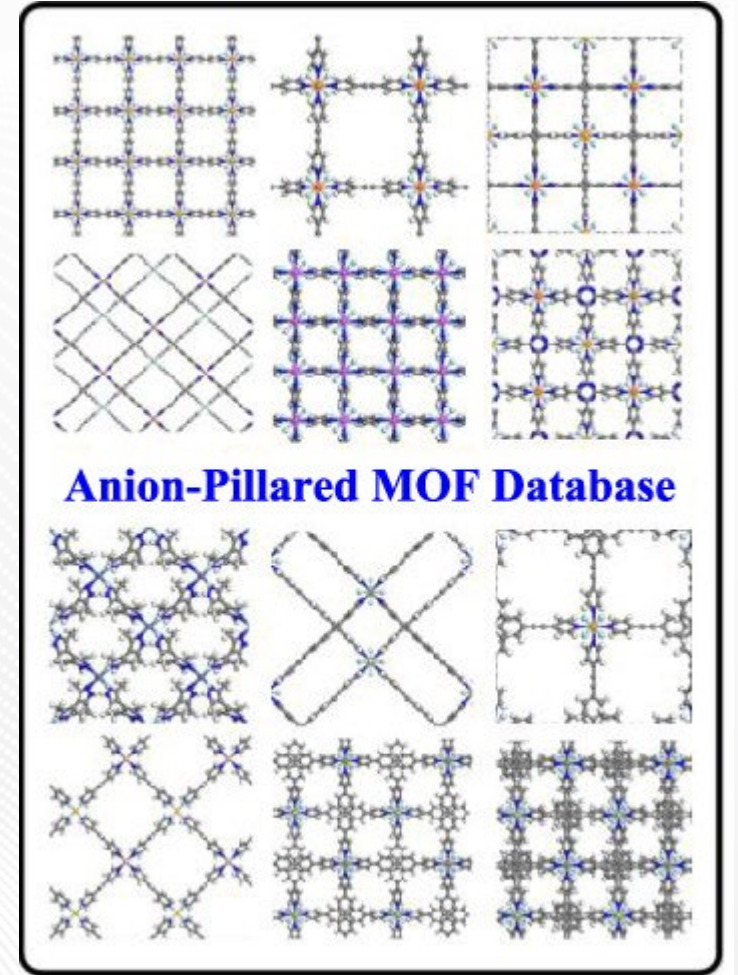
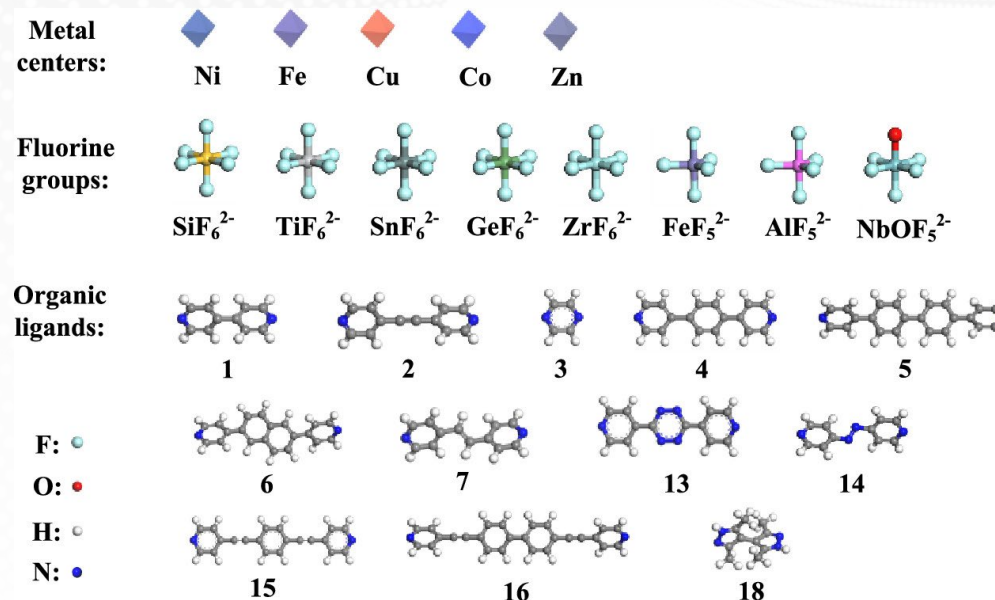


Lin et al., *Science* (2021).

Screening for Strong Physisorbing MOFs

Screening Anion-Pillared MOFs

- **Goal:** Identifying MOFs with high selectivity
 - Database from Gu and Sholl (2021)
 - Tunable metal center, fluorine groups, linkers
 - Systematic dataset: easy to obtain structure-property correlations



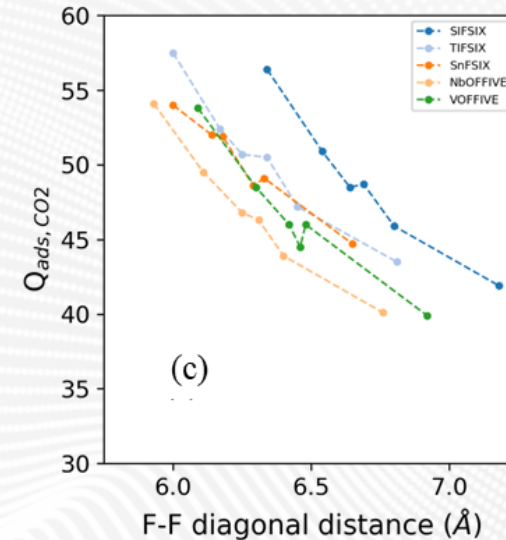
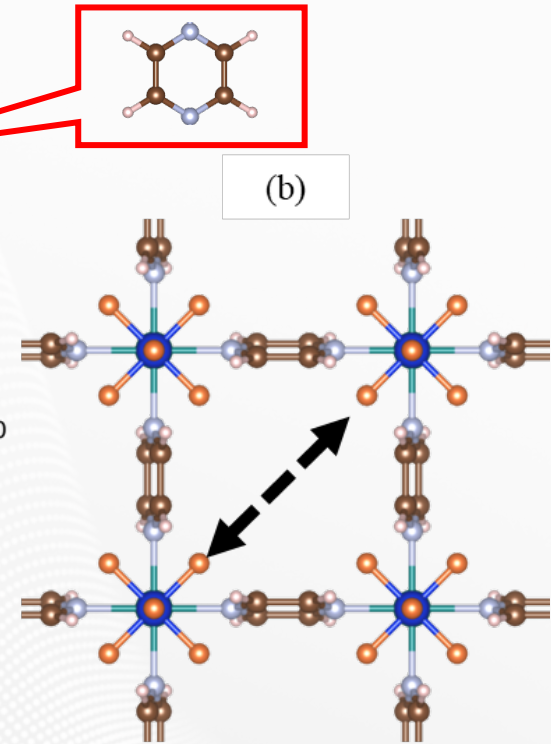
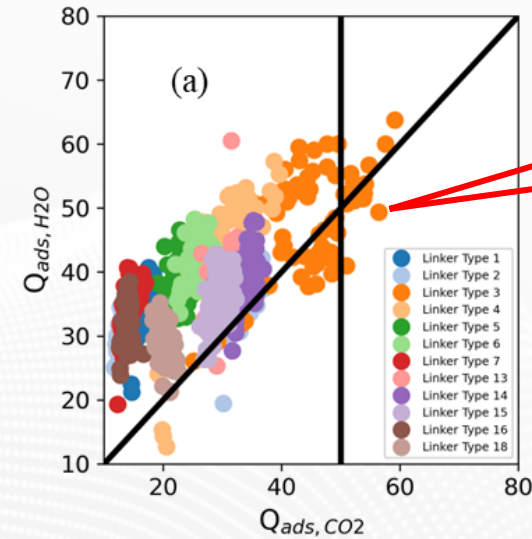
Screening for Physisorbent MOFs



MOFs with Pyrazine Linkers Show High Selectivity

- Screened database for MOFs based on zero-loading heats of adsorption
 - Pyrazine** shows high selectivity
 - Pyrazine** gives optimal F-F spacing

Fluorine Group	Linker	Metal	Q_{ads, CO_2} (kJ/mol)	Q_{ads, H_2O} (kJ/mol)
SIFSIX	Pyrazine	Cu	56.4	49.3
NbOFFIVE	Pyrazine	Cu	54.1	53.8
TIFSIX	Pyrazine	Ni	52.4	51.2
SIFIX	Pyrazine	Ni	50.9	41.0

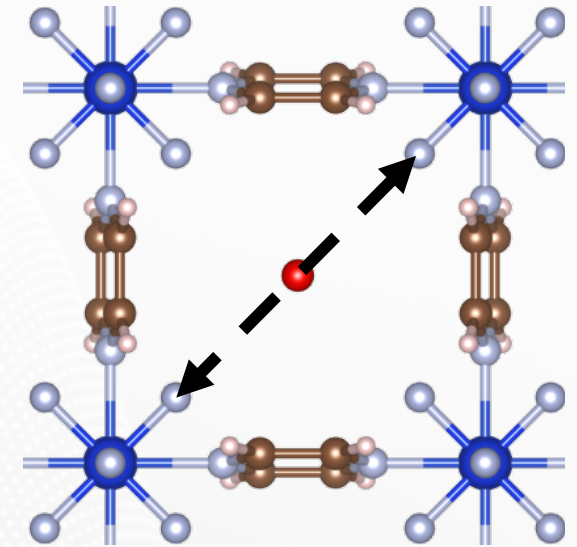


Findley et al. *J. Phys Chem C.* (2025).

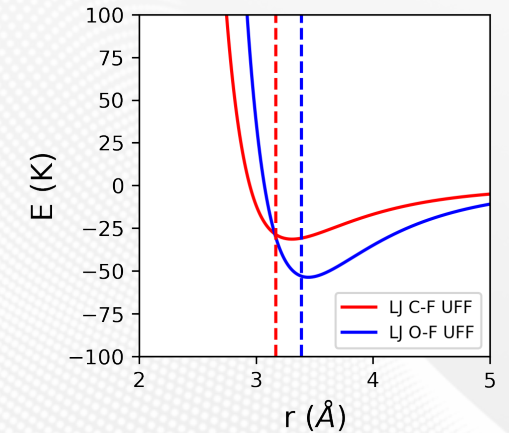
Importance of the Window

“Tug-of-War” Adsorption Sites

- Rigid, classical models showed strong adsorption in SIFSIX-3-Cu
 - Well-known material: good for benchmarking
- Adsorption site: Between four fluorine groups
- Correlation between window size and Q_{ads}^0
 - Optimal Van der Waals distances
- Flexibility of window can affect adsorption properties

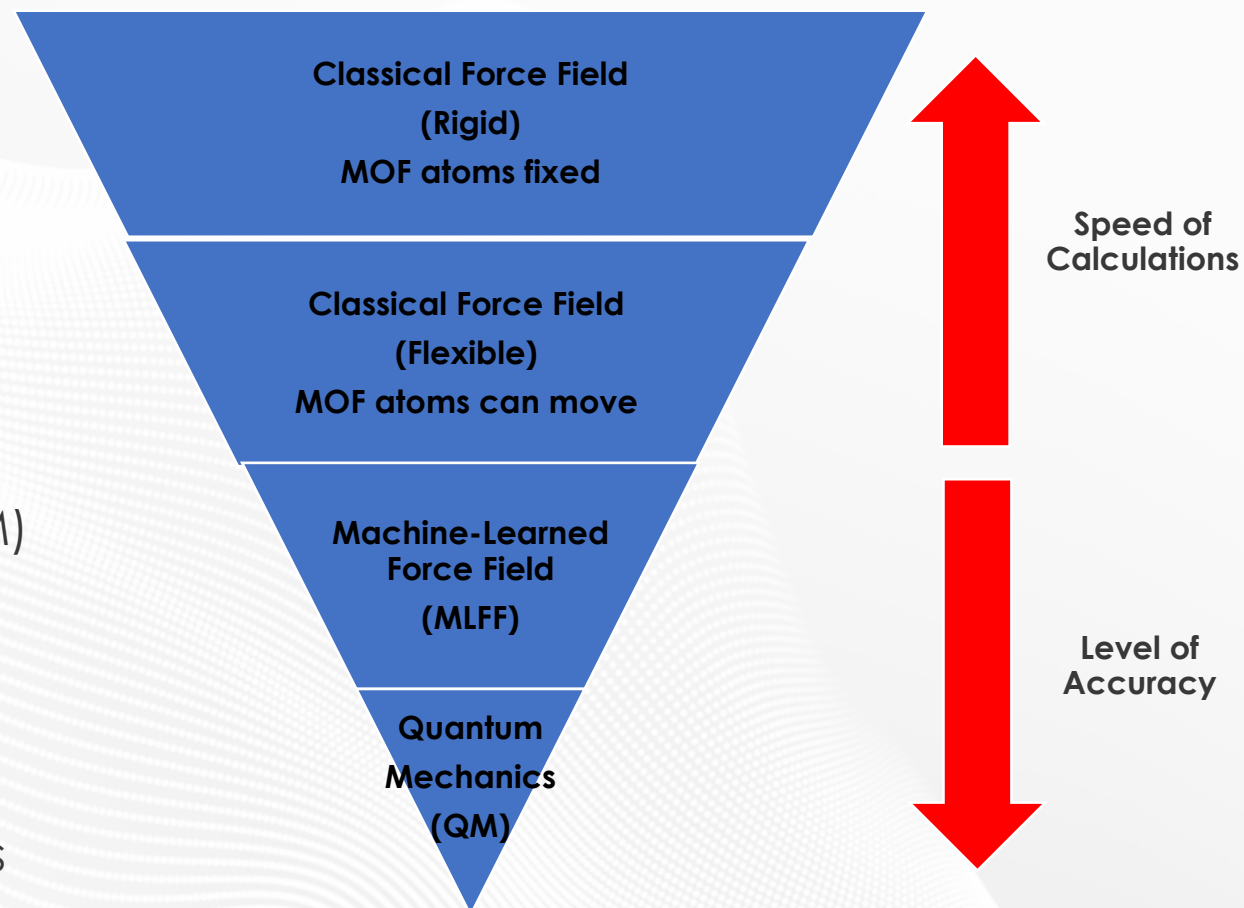


UFF Lennard-Jones



Machine-Learned Force Fields (MLFF) Connect Classical and Quantum Simulations

- **Classical Force Field (speed):**
 - **Rigid:** MOF atoms fixed during simulation
 - **Flexible:** MOF atoms can move
 - Energy, force from analytical function
 - Can calculate adsorption, diffusion
- **MLFF:**
 - Energy, force from ML model
 - Better fitting to quantum mechanical (QM) calculations
- **QM methods (accuracy):**
 - Such as density functional theory (DFT)
 - Accurate, slow
 - Compute energy, forces, determine structures

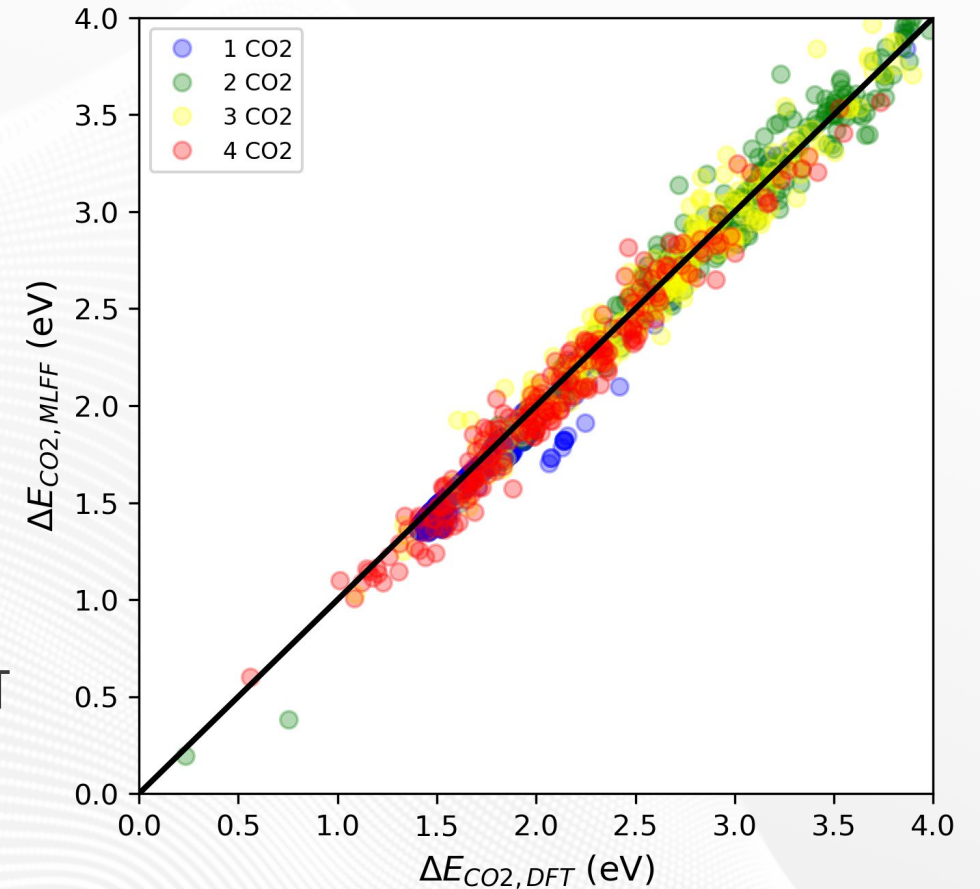


MLFF Fitting Procedure



Train → Simulate → Compare (to DFT) → Retrain & Repeat

- Initial Training data from DFT
 - Optimization
 - Ab initio molecular dynamics
 - Empty to fully saturated sorbent
- Fit MLFF (DeePMD)
 - Deep learning potential
 - $\Delta E = E_{\text{MOF+nads}} - E_{\text{MOF, opt}} - n_{\text{ads}} * E_{\text{ads}}$
- Molecular dynamics (MD) using MLFF
 - 1 ns trajectory
 - Loadings of empty to fully saturated
- Compute energy, force, stress of snapshots using DFT
- Retrain if $R^2 < 0.9$
 - Converged after two cycles



Generation of a Robust Dataset for Deep Learning

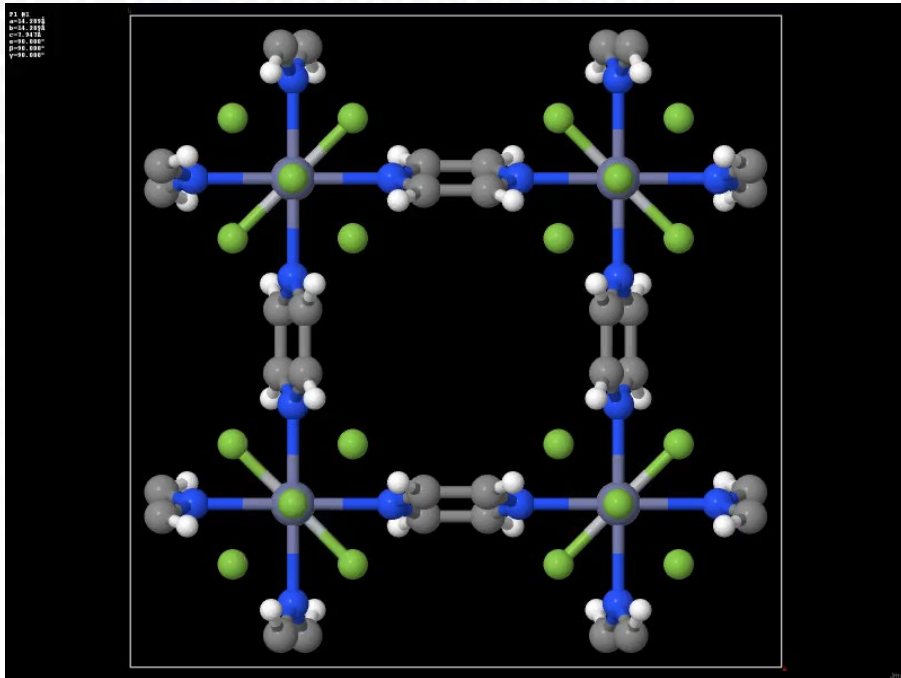


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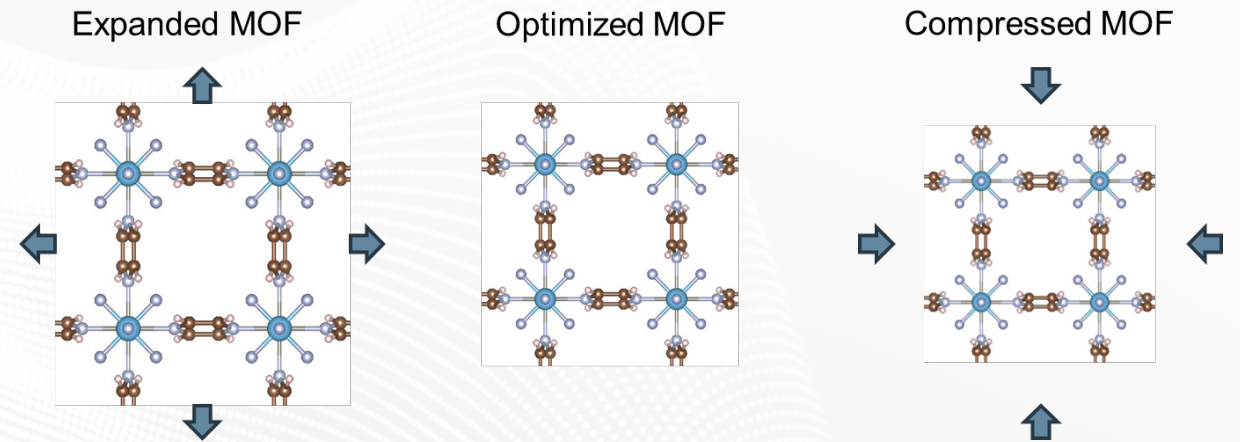


Generation of Initial Training Dataset

Sample Framework Dynamics
(both with and without adsorbates)

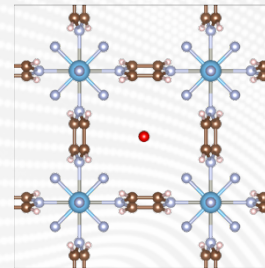


Sample Volume Changes

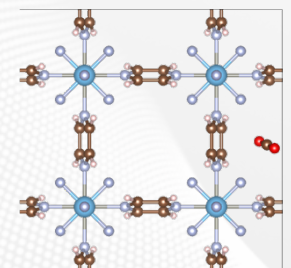


Sample CO₂ Adsorption

Lower Energy CO₂



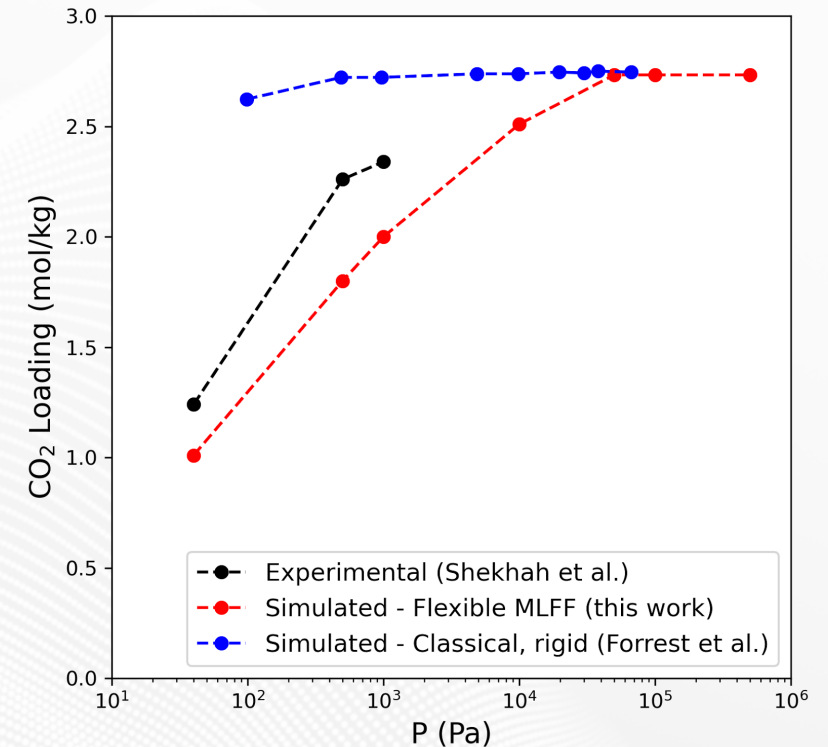
Higher Energy CO₂



Performance of MLFF Model

New Model Agrees Better with Experiments

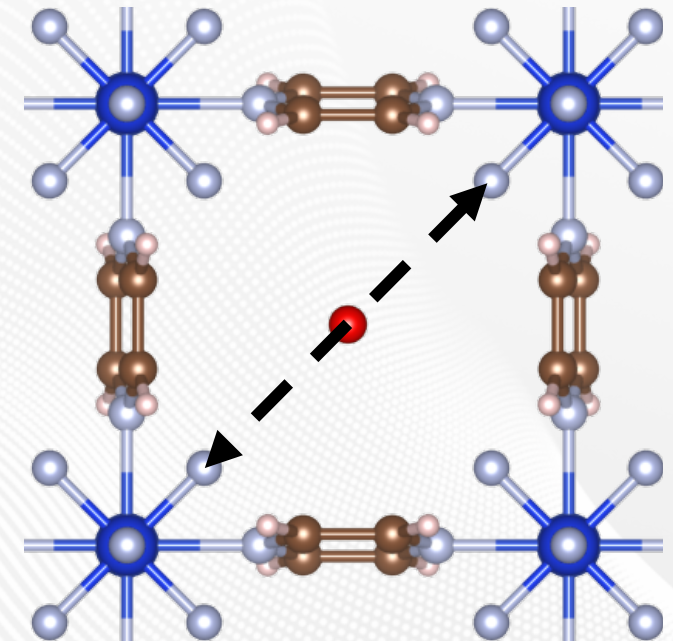
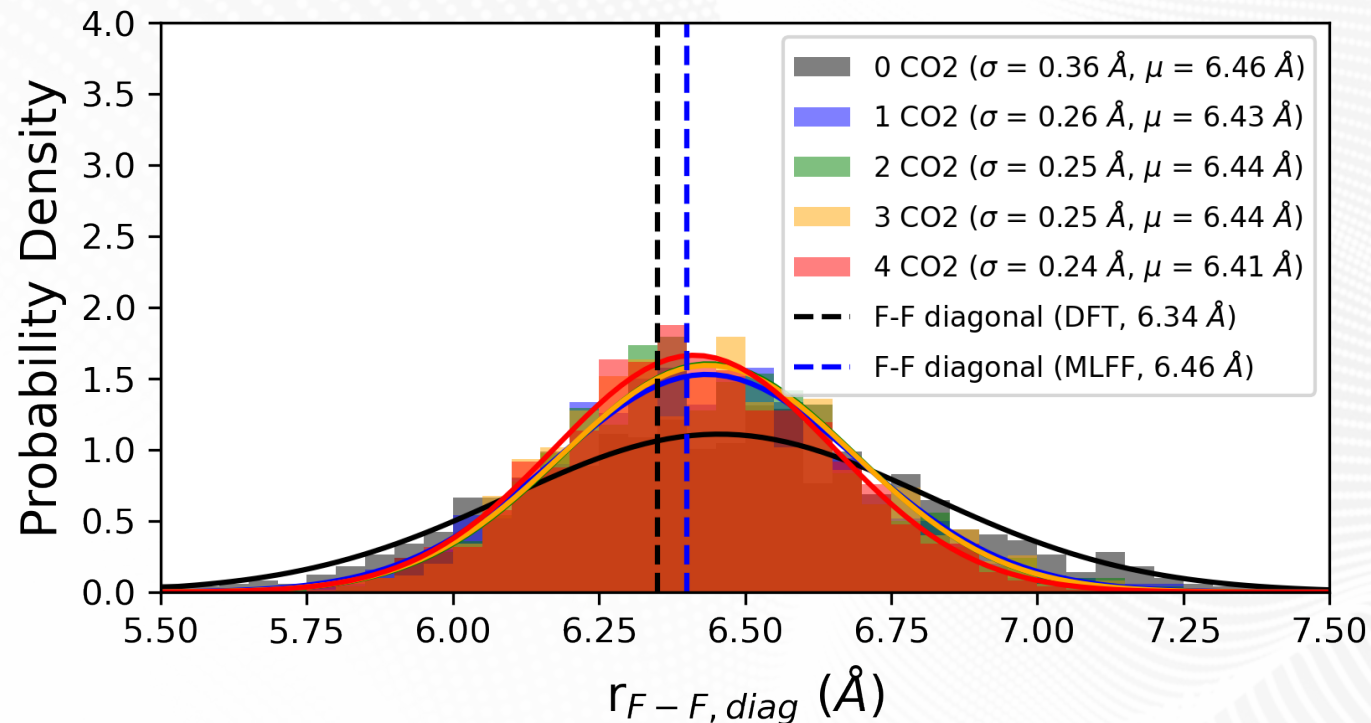
- MLFF used for hybrid Grand Canonical Monte Carlo (GCMC)/MD calculations
- Low pressure agreement improved significantly
- Why does framework flexibility have a larger effect at low pressure for this MOF?



Findley et al. *J. Phys Chem C.* (2025)
Shekhah et al. *Nature Communications* (2014)
Forrest et al. *Crystal Growth and Design* (2019)

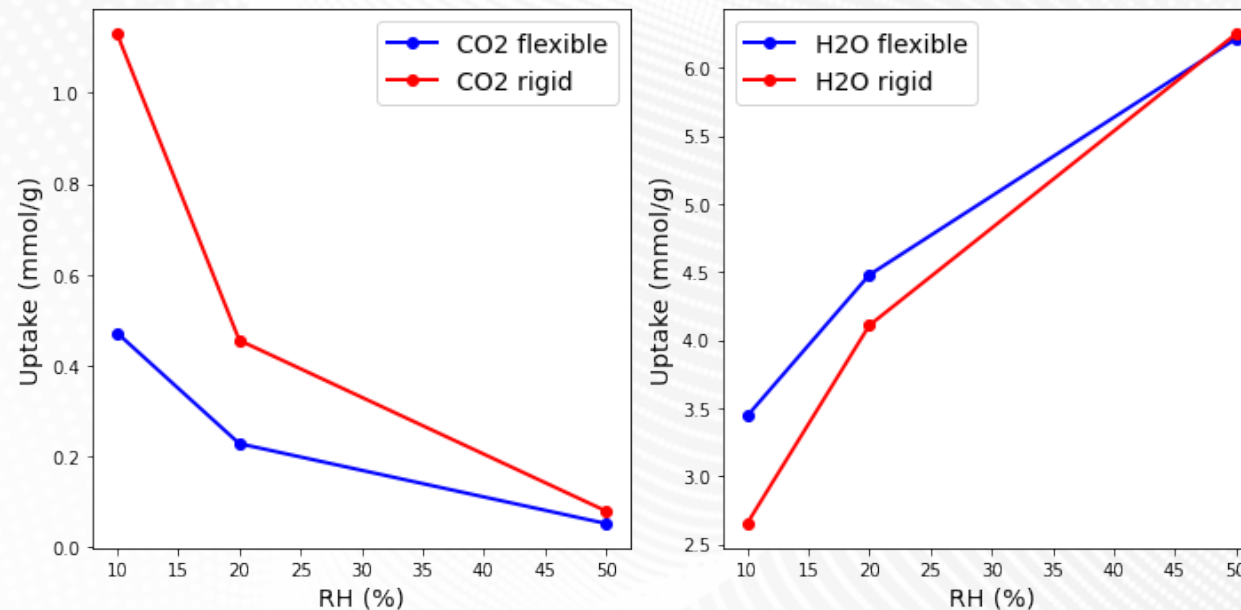
Window Size Varies Significantly (T = 300 K)

- MLFF used for MD at five different adsorbate loadings (0-4 molecules per unit cell)
- Strong adsorbate-framework interactions reduce framework flexibility
 - Reduction in flexibility is observed at loadings expected under industrial point source capture conditions
- Adding adsorbates slightly decreases average window size

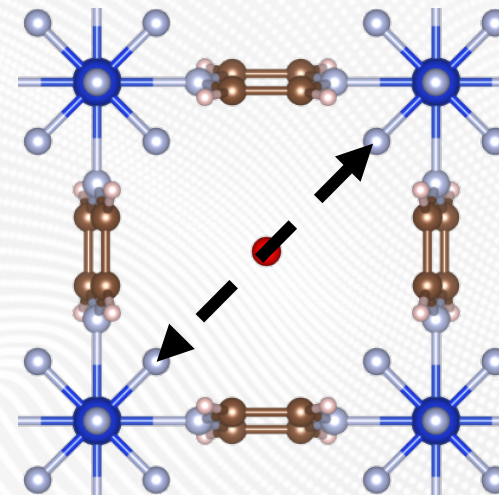
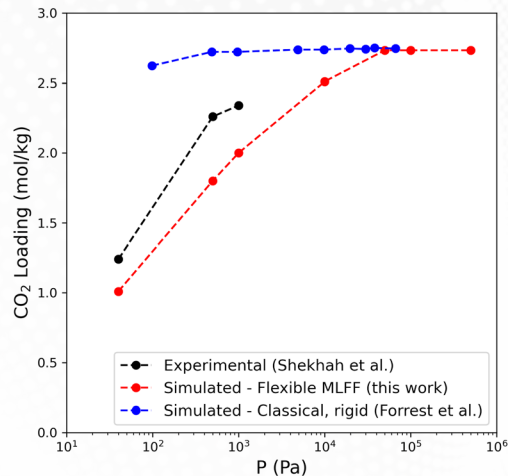


Binary GCMC of CO₂/H₂O Shows Competition

- Selective adsorption of CO₂ over H₂O assists in maintaining operational efficiency and reliability of gas processing units at power generation facilities
 - Accurate predictions of competitive adsorption result in more robust sorbent design
- Same images from MLFF-dynamics run
 - Two-component GCMC for each run, took average to get “flexible” approximation
- CO₂ and H₂O have similar affinities for SIFSIX-3-Cu, but different symmetries
 - Selectivity for CO₂ over H₂O decreases with flexibility, more prominent at low RH

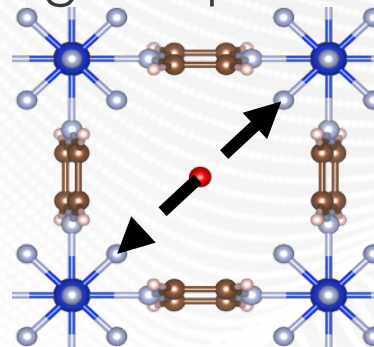


- Optimized material design through precise control of fluorine group spacing is critical for maximizing efficiency and throughput in industrial gas separation for reliable energy production.
- Machine Learning techniques significantly enhance our ability to predict material performance, accelerating the development of next-generation sorbents for American energy infrastructure.
- Understanding framework flexibility is key to designing robust, high-performance materials capable of sustained operation in demanding industrial environments associated with domestic energy production."



Thermal Conductivity, Water, Transferability

- **Thermal Conductivity:**
 - Quantifying the thermal management requirements to optimize energy efficiency and reduce operational costs in next-generation gas separation technologies for American industries
 - Does the presence of water or CO₂ affect thermal conductivity?
- **Water Adsorption:**
 - Investigating long-term material stability and performance under realistic operating conditions, especially with water, to ensure the reliability and economic viability of domestic energy utilization.
- **Transferability:**
 - Exploring synergistic impacts of flexibility and water on material performance to unlock new designs for robust and efficient gas separation in critical energy applications



Gu, Chenkai, et al. "Construction of an anion-pillared MOF database and the screening of MOFs suitable for Xe/Kr separation." *ACS Applied Materials & Interfaces* 13.9 (2021): 11039-11049.

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