

Development of Sorbents for Boron Capture from Coal Combustion Impoundment Leachates



Solutions for Today | Options for Tomorrow

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Authors



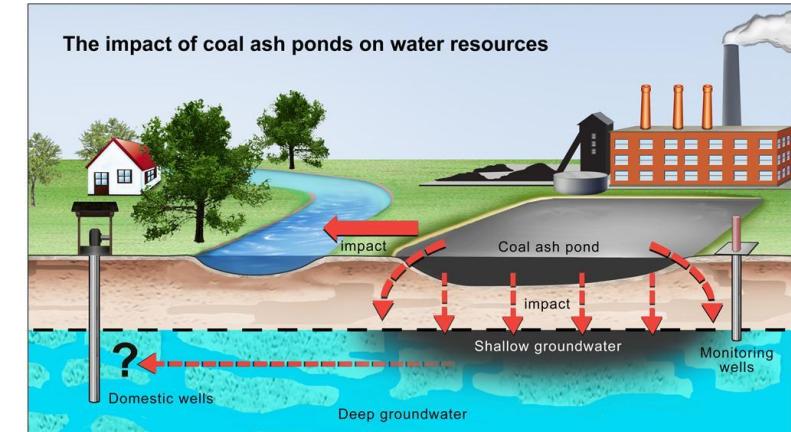
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Impoundment Sites

- Landfills to prevent the release of coal combustion products into the environment
 - Contain contaminants such as As, B, Hg, Se, Sr, etc.
- Boron: Present in impoundment sites
 - Primarily in the form of H_3BO_3 (acidic) or $B(OH)_4^-$ (basic)
- Zeolites: can be synthesized from fly ash
 - Have shown usefulness in removing contaminants
 - Topological variability
 - Tunable composition
- Which zeolite should be used for contaminant removal?

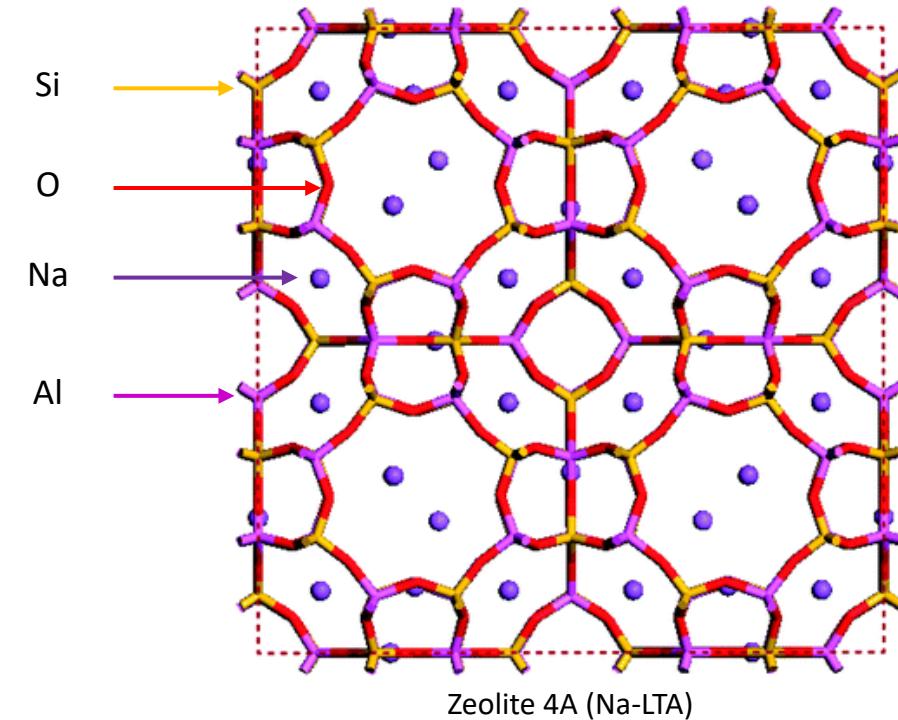


IZA Code	Alternate Names	PLD (Å)	LCD (Å)
ANA	Analcime	2.43	4.21
CHA	Chabazite Hershelite (Na-form) K-chabazite (K-form)	3.72	7.37
FAU	Faujasite Zeolite X ($Si/Al < 2$) Zeolite Y ($Si/Al \geq 2$)	7.35	11.2
GIS	Zeolite P1 NaP1	3.32	4.97
LTA	Zeolite 4A Zeolite 5A Zeolite A Linde Type A	4.21	11.05
LTF	Linde Type F	7.5 (z-direction)	8.16
LTL	Linde Type L Perlialite	7.5 (z-direction)	10.01
PHI	Phillipsite	3.69	5.40
SOD	Sodalite	2.53	6.32

Zeolites

Introduction

- Tetrahedral aluminosilicates
 - 245 distinct experimentally-synthesized topologies
 - Millions of hypothetical zeolites
- Composed of AlO_4 and SiO_4 tetrahedra
 - Substitution of Al for Si leads to charge imbalance
 - Extra-framework cations (Na^+ , K^+ , Ca^{2+} , etc.) balance charge
 - Cations are loosely bound, **can be exchanged**
 - Cations are adsorption and catalytic sites
 - Properties vary based on topology, composition, and Al distribution
- Uses for separations and catalysis
 - Stable, inexpensive to produce
 - Can be synthesized from coal fly ash
 - High internal surface area for adsorption

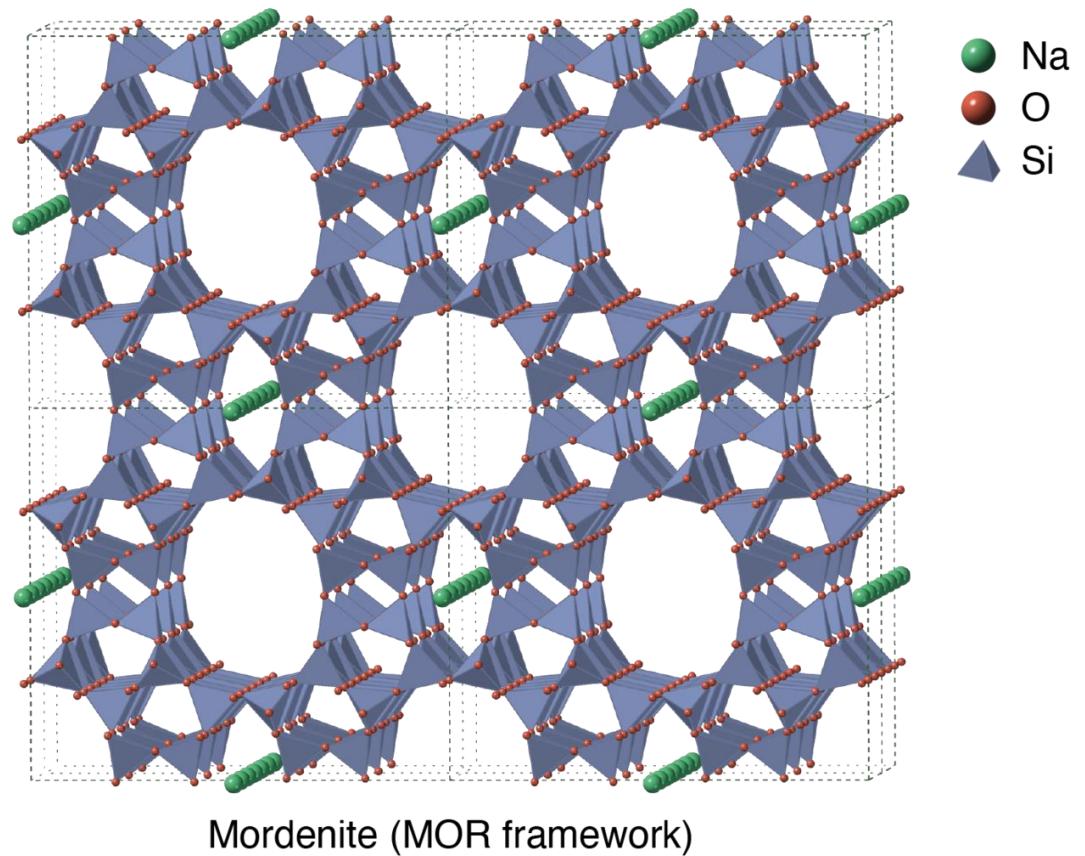


Number of Cations	Species of Cations	Zeolite Structure (Topology)
Si/Al ratio	Modified by cation exchange Ex. Na, K, Ca, Mn(II), Fe(II)	Controls pore size, shape, and surface area

Methodology

Overall Strategy

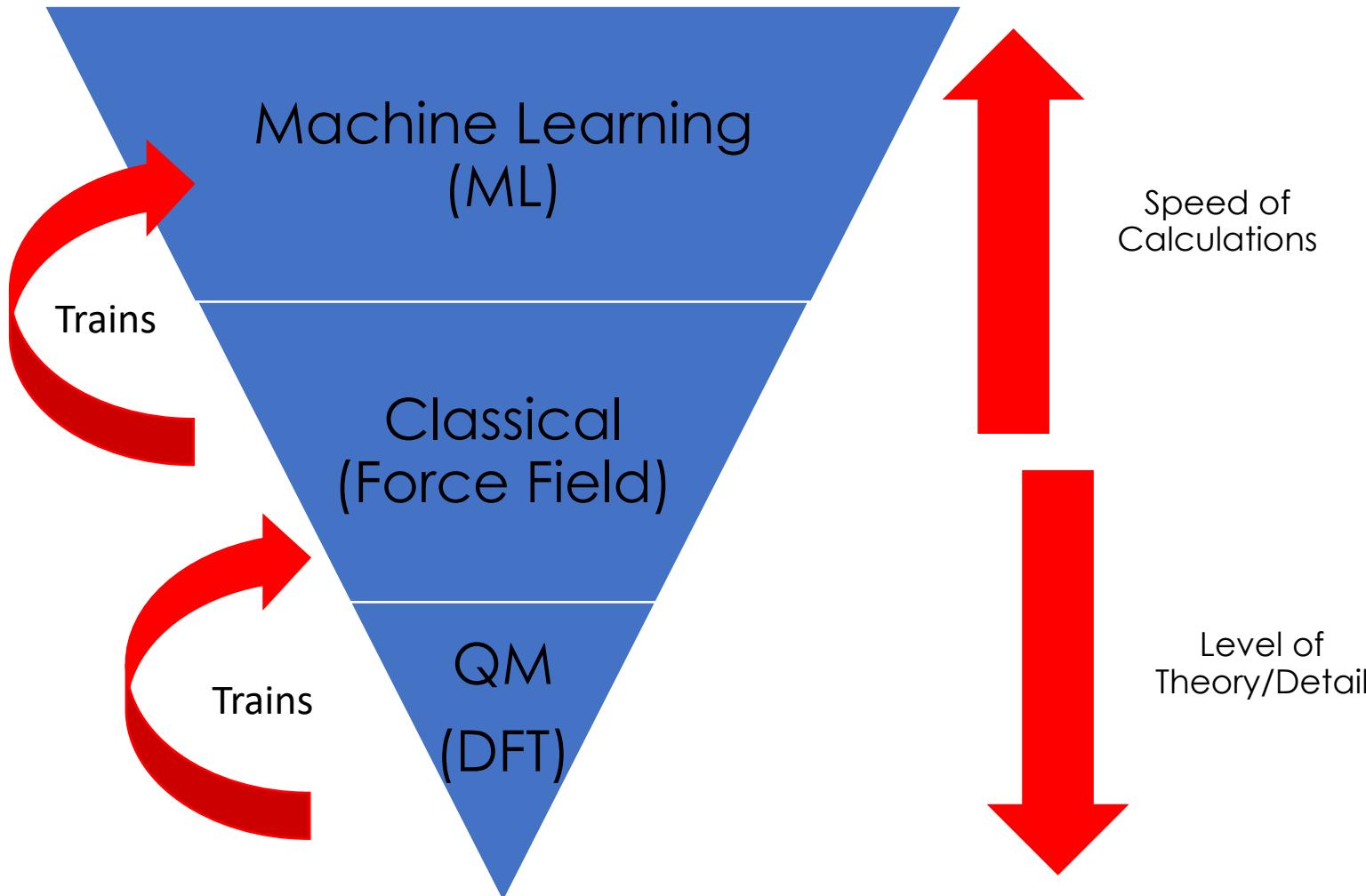
1. Construct a collection of sorbent structures and sorption conditions
2. Construct appropriate model potentials
3. Carry out computations to **estimate sorption** in a **representative subset** of the sorbent structures
4. Use AI/ML techniques to exploit relationships to **screen / design tailored sorbents** for impoundments



Machine Learning for Materials Design



Overall Strategy



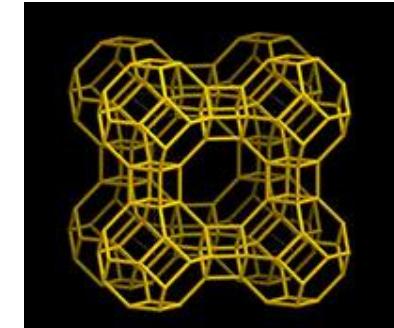
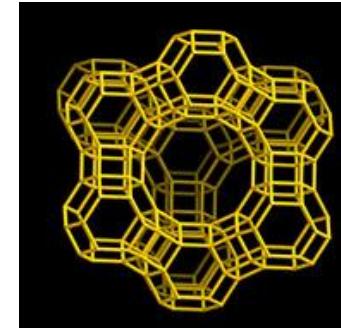
- **Machine Learning (ML):**
 - Accuracy depends on training set
 - Can be used for screening or design
 - Fastest of these three methods
- **Classical Simulations (FFs):**
 - Molecular simulations
 - Can study macroscopic properties
 - Useful for screening
- **Quantum Mechanical (QM) simulations:**
 - Using density functional theory (DFT) as the QM method
 - Accurate
 - Based on first-principles calculations
 - Cannot study macroscopic properties

Methodology

1. Construct a Collection of Sorbent Structures and Sorption Conditions

- **Framework symmetry:**
 - The framework symmetry controls geometrical features:
 - Pore-limiting diameters
 - Cavity diameter
 - Accessible surface area
 - Seven framework symmetries – zeolites with these symmetries have been synthesized from fly ash.
 - An additional five symmetries were chosen because they are common frameworks.
 - DDR, FER, MEL, MFI, TON
 - **Expanded list of framework symmetries: 12**

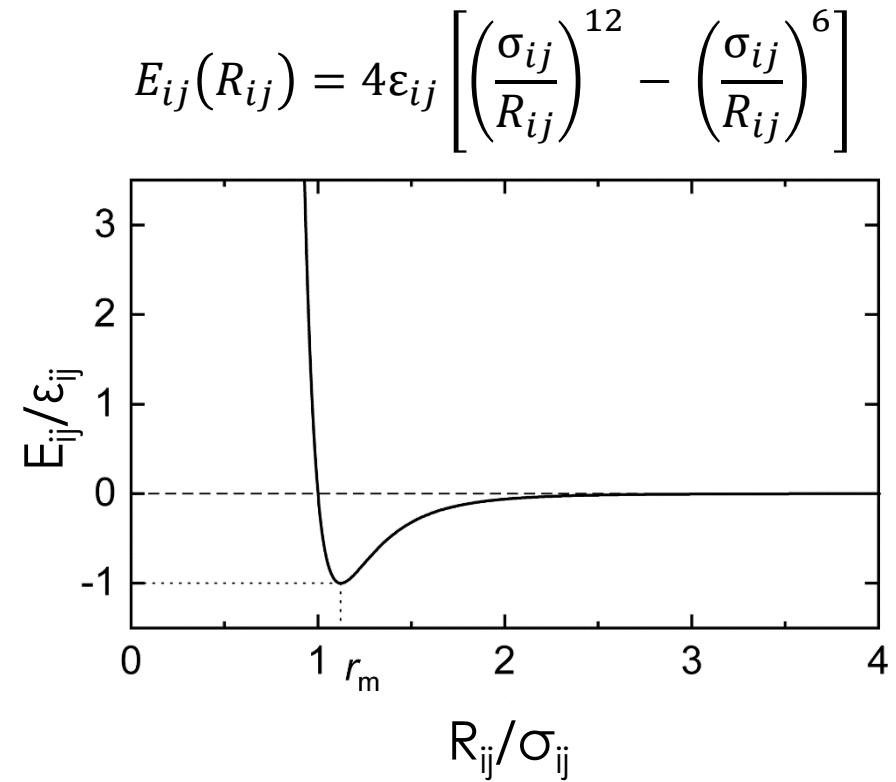
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Force Fields

2. Construct Appropriate Model Potentials

- Analytical function that describes interaction energies
 - Function of distance between a pair of atoms (R_{ij})
 - Total energy of system is the sum over the atomic pairs
 - **Example:** Lennard-Jones potential (right)
 - σ is related to average atomic size
 - ϵ is related to depth of potential energy well
- Used with statistical mechanics to calculate thermodynamic properties
 - Fast computation of energies means more configurations and better statistics
 - Phase equilibrium, heats of adsorption, adsorption isotherms
- Often parametrized based on **experimental data** or **QM calculations (DFT)**

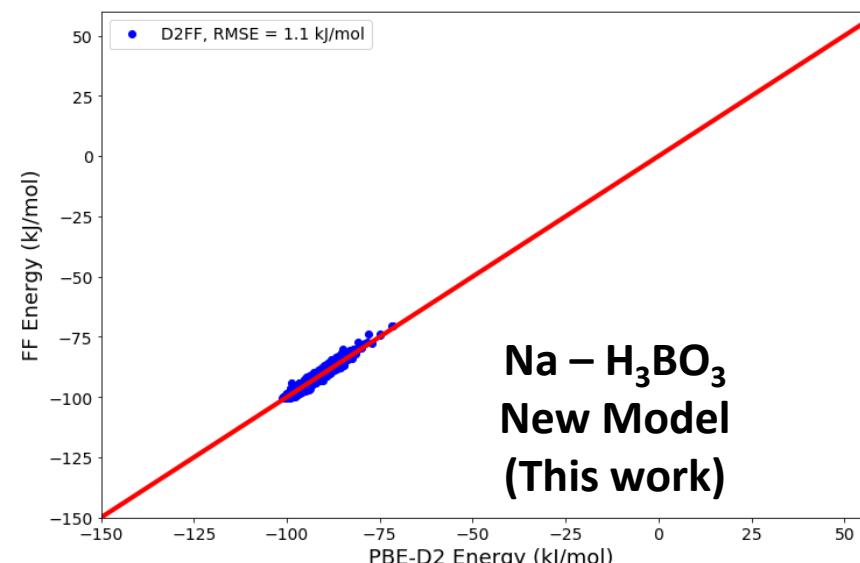
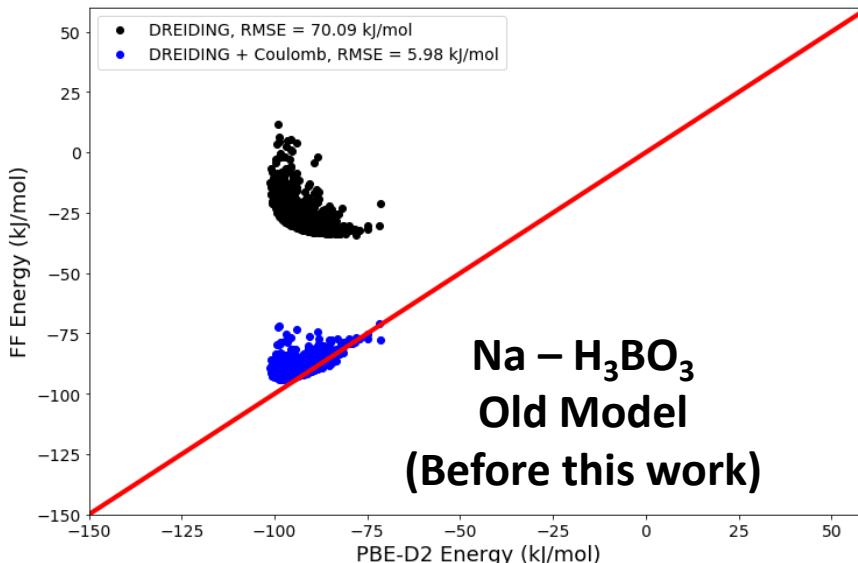


Fitting Adsorbate – Zeolite Interactions

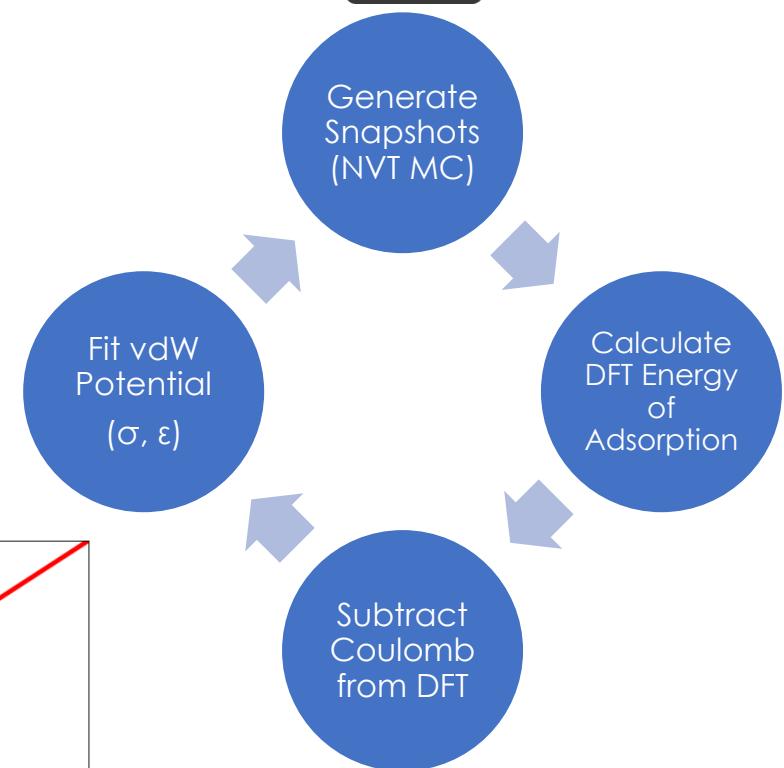


2. Construct Appropriate Model Potentials

- DFT optimization showed **no chemical bonds broke** during adsorption of H_3BO_3 , H_2O
 - Can use classical force fields for adsorption
- Complete for Na^+ , K^+ , Ca^{2+} , Mn^{2+} , Fe^{2+}
- Construct models to reproduce adsorption energies from DFT



$$E_{ij}(R_{ij}) = 4\epsilon_{ij} \left[\left(\frac{\sigma_{ij}}{R_{ij}} \right)^{12} - \left(\frac{\sigma_{ij}}{R_{ij}} \right)^6 \right] + \frac{q_i q_j}{R_{ij}}$$

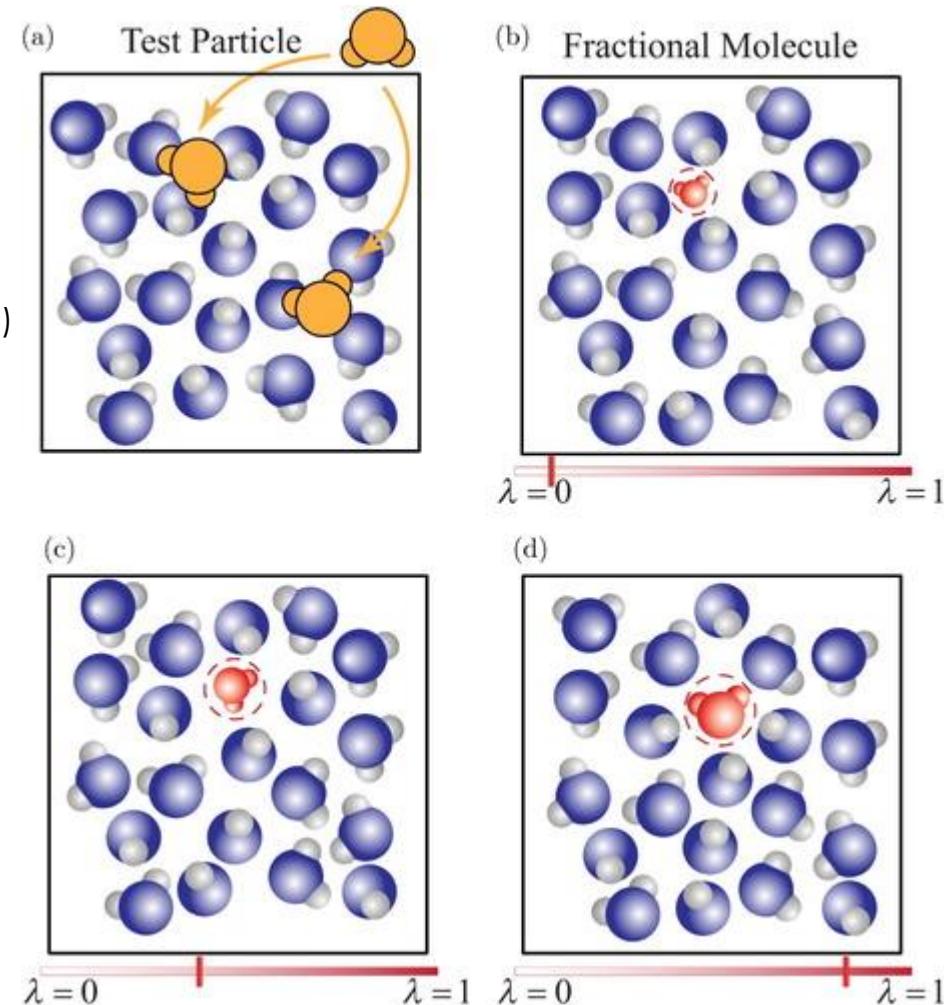


Adsorption Simulation Methods

3. Carry out Computations to Estimate Sorption in Sorbent Structures

- Adsorption from solution:
 - Reference state is now boric acid in solution
 - $f_A = \rho k_B T \left(e^{\mu_A^\infty / k_B T} \right) x_A$
 - μ_A^∞ is the excess chemical potential of infinitely dilute solute in the solvent
 - x_A is mole fraction of solute
 - Model predictions for Henry's constant match Hazardous Substances Data Bank (HSDB)
 - $K_H = \lim_{x_A \rightarrow 0} \left(\frac{f_A}{x_A} \right) = \rho k_B T \exp(\beta \mu_A^\infty)$
- Continuous Fractional Component Monte Carlo
 - Designed for dense systems: Difficult to perform insertion moves without overlap
 - Insertion of boric acid in hydrophilic zeolite
 - Inserts a "fractional" molecule
 - Scales intermolecular interactions by λ (0, 1]
 - Allows other molecules to move with fractional molecule present
 - Increases weight (λ) until full molecule is grown

	K_H (Pa*m ³ /mol)	Source
Experimental	2.65×10^{-7}	HSDB (NIH)
Sim (CFCMC-WI)	1.29×10^{-7}	This work



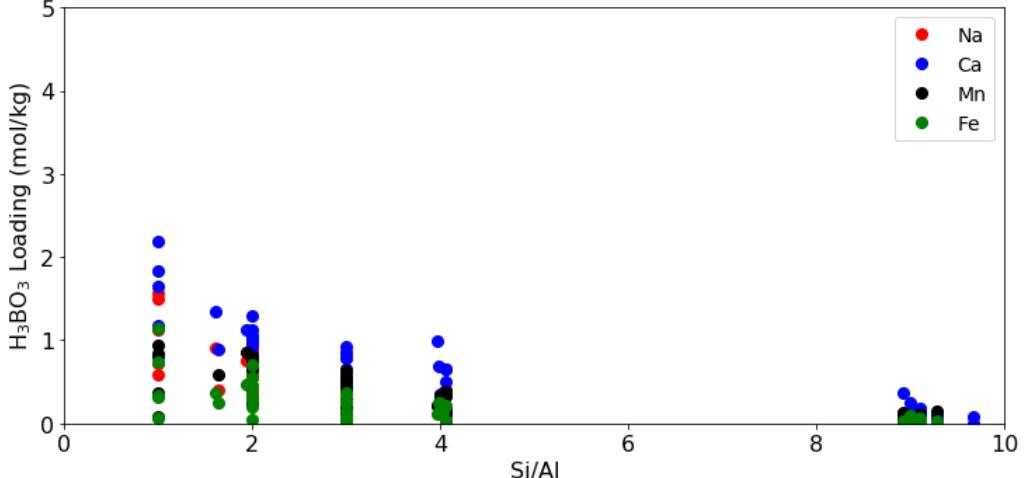
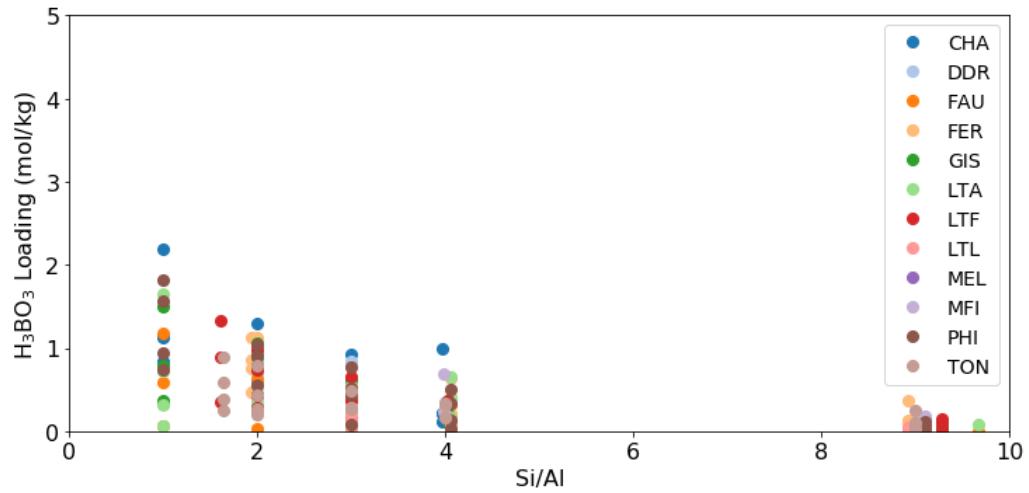
Rahbari, A., et al. "Recent advances in the continuous fractional component Monte Carlo methodology." *Molecular Simulation* 47.10-11 (2021): 804-823.

Xiong et al. Alcohol Adsorption onto Silicalite from Aqueous Solution *Journal of Phys. Chem. C* (2011)

Initial Dataset (1 ppm)

3. Carry out Computations to Estimate Sorption in Sorbent Structures

- Results for 1 ppm H_3BO_3 solution
- Ca, Na-exchanged zeolites were top performers
 - Largest charge (Ca)
 - Most cations per unit cell (Na)
- Low Si/Al ratios performed best
 - Most cations per unit cell

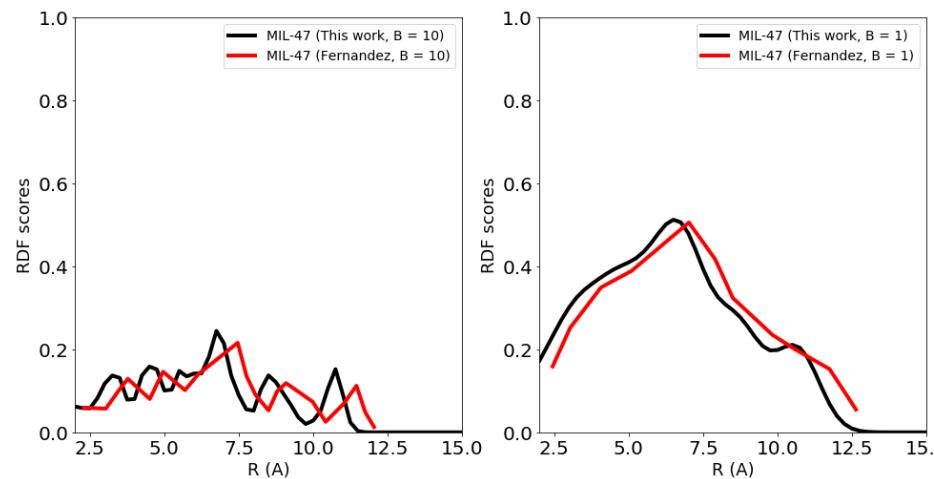


Zeolite Structure Encoding (RDFs)

4. Use AI/ML Techniques to Exploit Relationships to Design Tailored Sorbents

Atomic Property Weighted Radial Distribution Functions

- RDF analysis is a crystallographic technique sensitive to both short- and long-range structural correlations.
- The RDF is the interatomic separation histogram representing the weighted probability of finding a pair of atoms separated by a given distance.
- The RDFs can be weighted to fit the requirements of the chemical information to be represented, by introducing the atomic properties, P_i .
- Electronegativity, polarizability, and van der Waals volume.
- Encoded RDFs for all IZA zeolites.



$$\text{RDF}^P(R) = f \sum_{i,j}^{\text{all atom pairs}} P_i P_j e^{-B(r_{ij}-R)^2}$$

Fernandez, M.; Trefiak, N. R.; Woo, T. K. Atomic Property Weighted Radial Distribution Functions Descriptors of Cation–Organic Frameworks for the Prediction of Gas Uptake Capacity. *J. Phys. Chem. C* **2013**, 117 (27), 14095–14105.

Machine Learning Model



4. Use AI/ML Techniques to Exploit Relationships to Design Tailored Sorbents

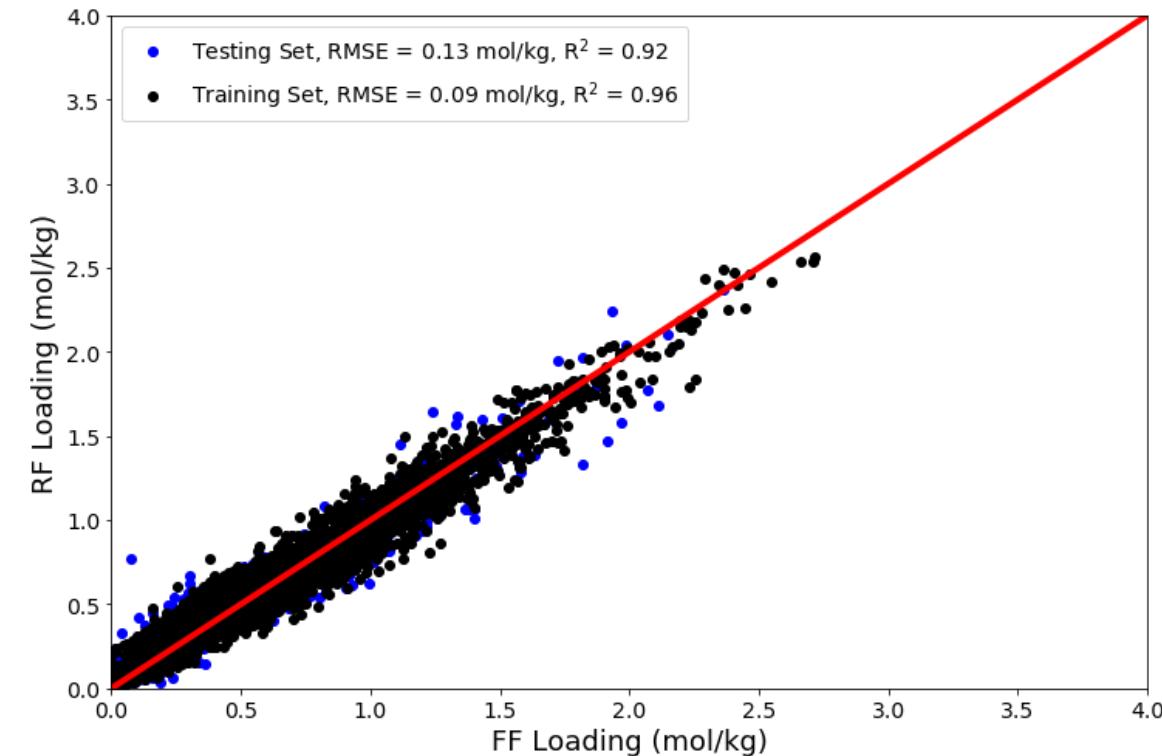
- ML Model requires:
 - **Descriptors:**
 - Crystal structure (topology)
 - Composition (both zeolite and solution)
 - **Adsorption data:**
 - Little experimental data available (not enough for ML)
 - Can predict adsorption using molecular simulations
 - Accuracy depends on models and level of theory
 - Can use molecular simulations to generate a large, robust training set

Target	Number of Cations	Species of Cations				Zeolite Structure Encoding		Solution
Amount Adsorbed	Al/O	Na/O	Ca/O	Mn/O	Fe/O	RDF_PC1	RDF_PC2	Boric Acid Concentration

$$y = F(x_1, x_2, \dots, x_n)$$

4. Use AI/ML Techniques to Exploit Relationships to Design Tailored Sorbents

- **Prediction:** H_3BO_3 uptake from 1-20 ppm H_3BO_3 solution
- **Features:**
 - Stoichiometry (normalized by number of O atoms)
 - Six principal components for weighted RDFs (weighted by charge, electronegativity)
 - H_3BO_3 concentration in solution
- **Model:** Random Forest Regression
 - Max depth = 10
 - N trees = 200
 - Tuned using gridsearchCV (5-fold)
- **Conclusions:**
 - Good quality of fit
 - Most important features: AlO (Si/Al ratio), CaO (cation type)



Genetic Algorithms – Zeolite Optimization



4. Use AI/ML Techniques to Exploit Relationships to Design Tailored Sorbents

- **“Survival of the fittest”**
 - Optimization technique
- **Genes:** Features
 - Examples. Si/Al, Al/O, RDF_PC1
- **Chromosome:** Set of genes (a zeolite)
 - Example: Na-LTA with Si/Al = 1
- **Population:** Group of potential solutions
- **Fitness function:** Affects the probability of selection for “reproduction”
 - Predicted H_3BO_3 uptake
- **Crossover:** Swapping a set of genes between two chromosomes and adding the offspring to the population
- **Mutation:** Altering the set of genes in an offspring

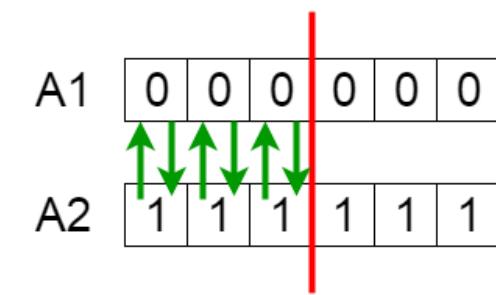
Gene	Values	Meaning
Topology	[0 to $N_{\text{topologies}}$]	Assigns structure encoding
AlO	[0, 1]	Si/Al ratio
NaO	[0, 1]	Na%
CaO	[0, 1]	Ca%
MnO	[0, 1]	Mn%
FeO	[0, 1]	Fe%

Before Mutation

A5 1 1 1 0 0 0

After Mutation

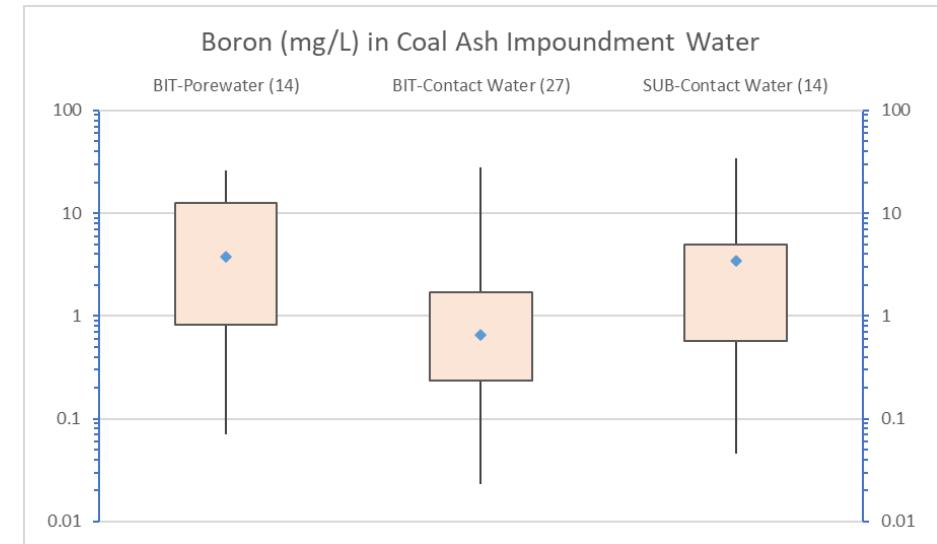
A5 1 1 0 1 1 0



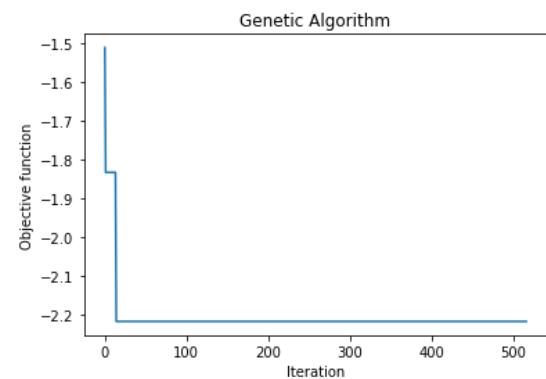
Genetic Algorithms – Zeolite Optimization

4. Use AI/ML Techniques to Exploit Relationships to Design Tailored Sorbents

- Selected boron concentration data from Electric Power Research Institute (EPRI) report
- Designed tailored zeolites using GA optimization
 - Case 1:** Fly ash zeolite topologies, all Si/Al are valid
 - Broader search
 - Case 2:** Fly ash zeolite topologies, minimum Si/Al based on literature search
 - Searches zeolites that are “easier” to make
- CHA and LTA topologies are ideal for boric acid removal
 - Si/Al = 1, partially Ca-exchanged



Type	Percentile	Concentration (mg/L)	Case	Loading (mol/kg)	Topology	Si/Al	Na (%)	Ca (%)	Mn (%)	Fe (%)
BIT-Porewater	75	12.84	Case 1	2.5	CHA	1	42.82	49.98	2.29	4.92
			Case 2	2.25	LTA	1	7.37	70.09	6.60	15.94
BIT-Contact Water	100	27.43	Case 1	2.5	CHA	1	49.27	46.25	3.24	1.24
			Case 2	2.41	LTA	1	5.75	74.12	3.40	16.73
SUB-Contact Water	50	3.50	Case 1	2.5	CHA	1	29.47	68.49	0.39	1.66
			Case 2	2.04	LTA	1	3.02	78.60	4.75	13.63



Next Steps



- **Model enhancement**
 - Model pH effects
- **Add additional contaminants to models (such as As, Hg, Ni, Se)**
 - Hg, Ni: ion-exchange
 - As, Se: adsorption

Questions?



NETL Resources

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CONTACT:

John Findley

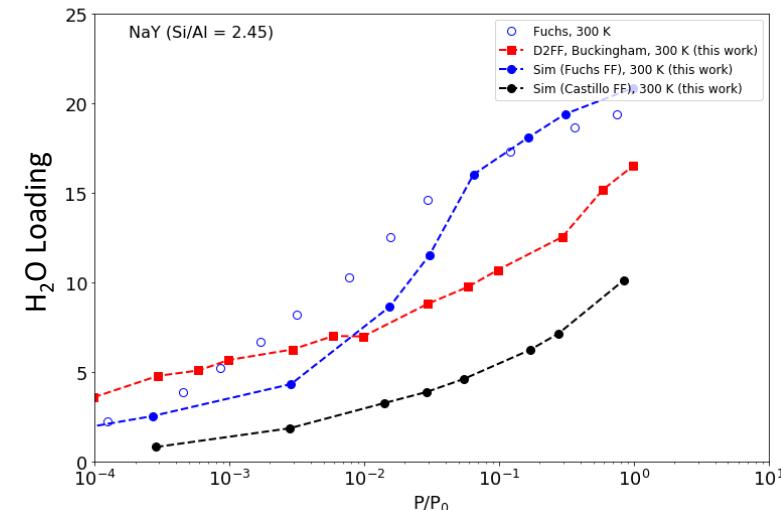
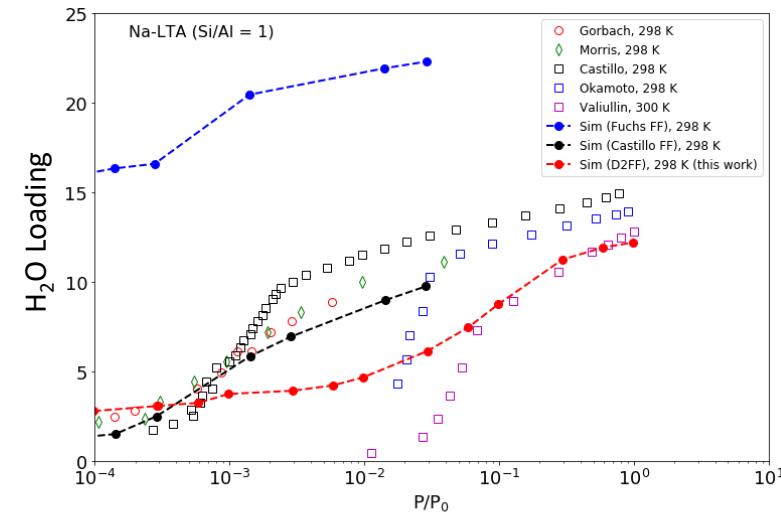
John.Findley@netl.doe.gov



Model Validation

2. Construct Appropriate Model Potentials (Validation)

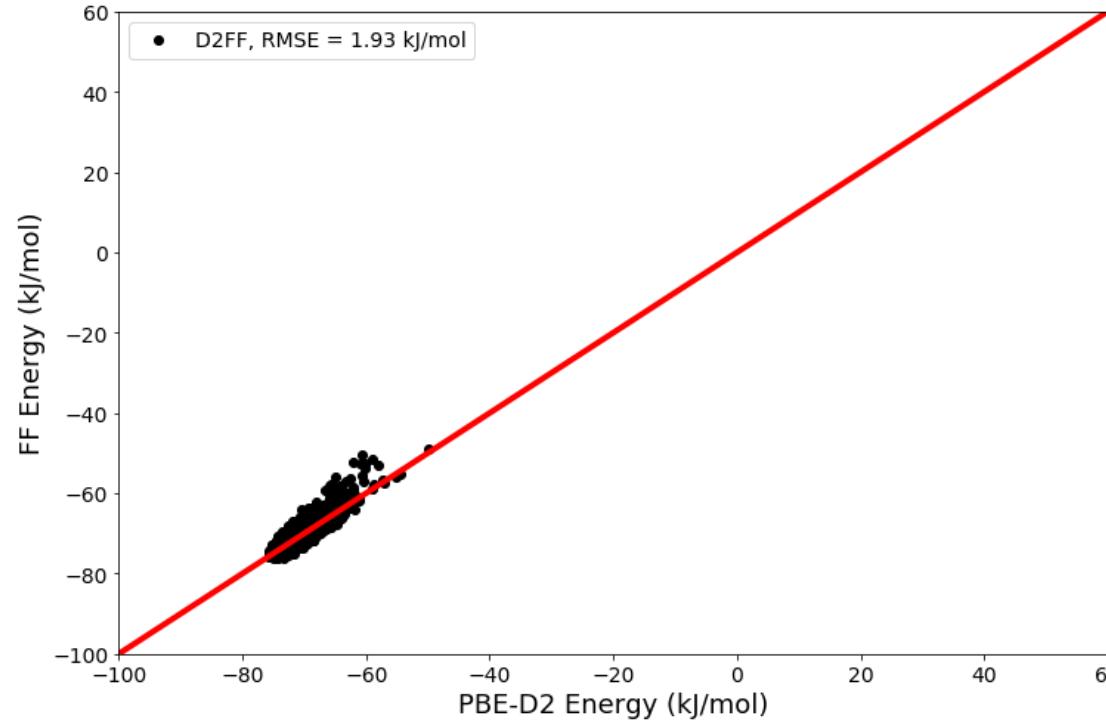
- Used models developed in December to predict water vapor adsorption isotherms
 - Model parametrized by fitting to Density Functional Theory Energies (PBE-D2 to be specific)
 - Compared two different potentials
 - Buckingham:** $E_{ij} = A_{ij}e^{-B_{ij}R_{ij}} - \frac{c_{ij}}{R_{ij}^6} + \frac{q_i q_j}{4\pi\epsilon_0 R_{ij}}$
 - Na-LTA (Si/Al = 1, top) and NaY (Si/Al = 2.45, bottom)
- Most important ranges are:
 - High pressure** (similar to liquid water)
 - Low pressure** (water-zeolite interaction strength)
- Low pressures:** Good agreement with **D2FF**
 - Our Buckingham model predicts interaction energies between the zeolite and water
 - This bodes well for our H_3BO_3 models because the fugacity of H_3BO_3 is very low for ppm-level concentrations
- High pressure (near saturation):** Good agreement
 - Similar to liquid water inside the pores
- Intermediate pressures:** Decent to poor agreement for **D2FF**
 - Shape of the isotherm does not completely match
 - Likely caused by imperfections in SPC/E water model (non-polarizable, only three-point charges)
- Our Model (D2FF) seems sufficiently accurate for further use**
 - Note:** Simulations (D2FF, red) performed **without** experimental input
 - Fuchs FF (blue) was fit to Fuchs experimental data (NaY)
 - Castillo FF (black) was fit to Castillo experimental data (black)



Addition of K

2. Construct appropriate model potentials

- Fit new models (when necessary) K interactions with zeolite framework (finished), H₂O (finished), H₃BO₃ (finished, pictured)



Atomic Pair	A _{ij} (K)	B _{ij} (Å ⁻¹)	C _{ij} (K*Å ⁶)	q _M (-e)	Source
K – Oz	6.23*10 ⁷	3.43	2.29*10 ⁶	0.990	Fang et al.
K – O (H ₂ O)	3.50*10 ⁶	2.55	0	0.990	This work

Atomic Pair	ε _{ij} (K)	σ _{ij} (Å)	q _M (-e)
K – O (H ₃ BO ₃)	348	2.87	0.990