

# Development of Sorbents for Boron Capture from Coal Combustion Impoundment Leachates



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**Solutions for Today | Options for Tomorrow**

The background of the slide is a repeating pattern of a complex molecular structure, likely a zeolite or a similar crystalline material, rendered in a 3D ball-and-stick model. The atoms are colored: carbon is grey, oxygen is red, nitrogen is blue, and hydrogen is white. The structure is set against a light blue background.

2022 Pittsburgh Coal Conference  
Sept. 20, 2022

# Disclaimer



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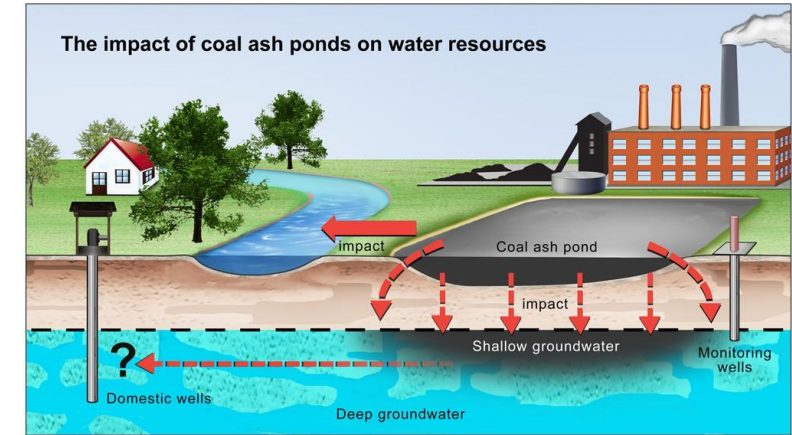
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# Impoundment Sites

- 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  $\text{H}_3\text{BO}_3$  (acidic) or  $\text{B}(\text{OH})_4^-$  (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?

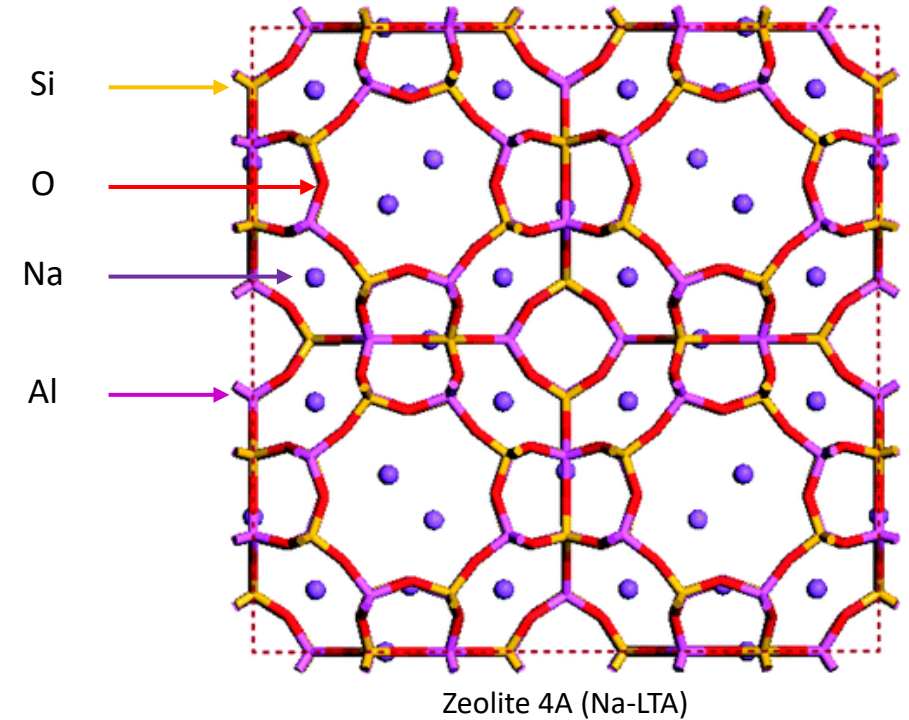


IZA Code	Alternate Names	PLD (Å)	LCD (Å)
ANA	Analcime	2.43	4.21
CHA	Chabazite Hershelite (Na-form) K-chabazite (K-form)	3.72	7.37
FAU	Faujasite Zeolite X (Si/Al < 2) Zeolite Y (Si/Al >= 2)	7.35	11.2
GIS	Zeolite P1 NaP1	3.32	4.97
LTA	Zeolite 4A Zeolite 5A Zeolite A Linde Type A	4.21	11.05
LTF	Linde Type F	7.5 (z-direction)	8.16
LTL	Linde Type L Perliaite	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  $\text{AlO}_4$  and  $\text{SiO}_4$  tetrahedra
  - Substitution of Al for Si leads to charge imbalance
  - Extra-framework cations ( $\text{Na}^+$ ,  $\text{K}^+$ ,  $\text{Ca}^{2+}$ , 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



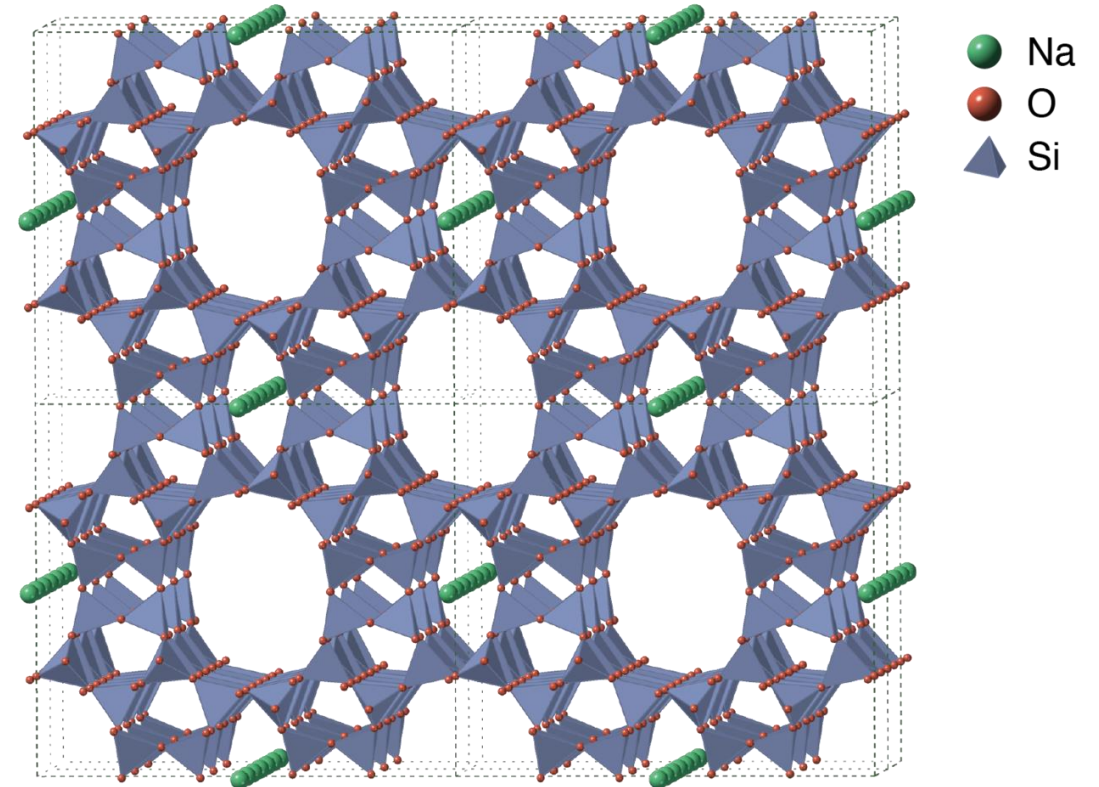
Number of Cations	Species of Cations	Zeolite Structure (Topology)
Si/Al ratio	Modified by cation exchange Ex. Na, K, Ca, Mn(II), Fe(II)	Controls pore size, shape, and surface area



# Methodology

## Overall Strategy

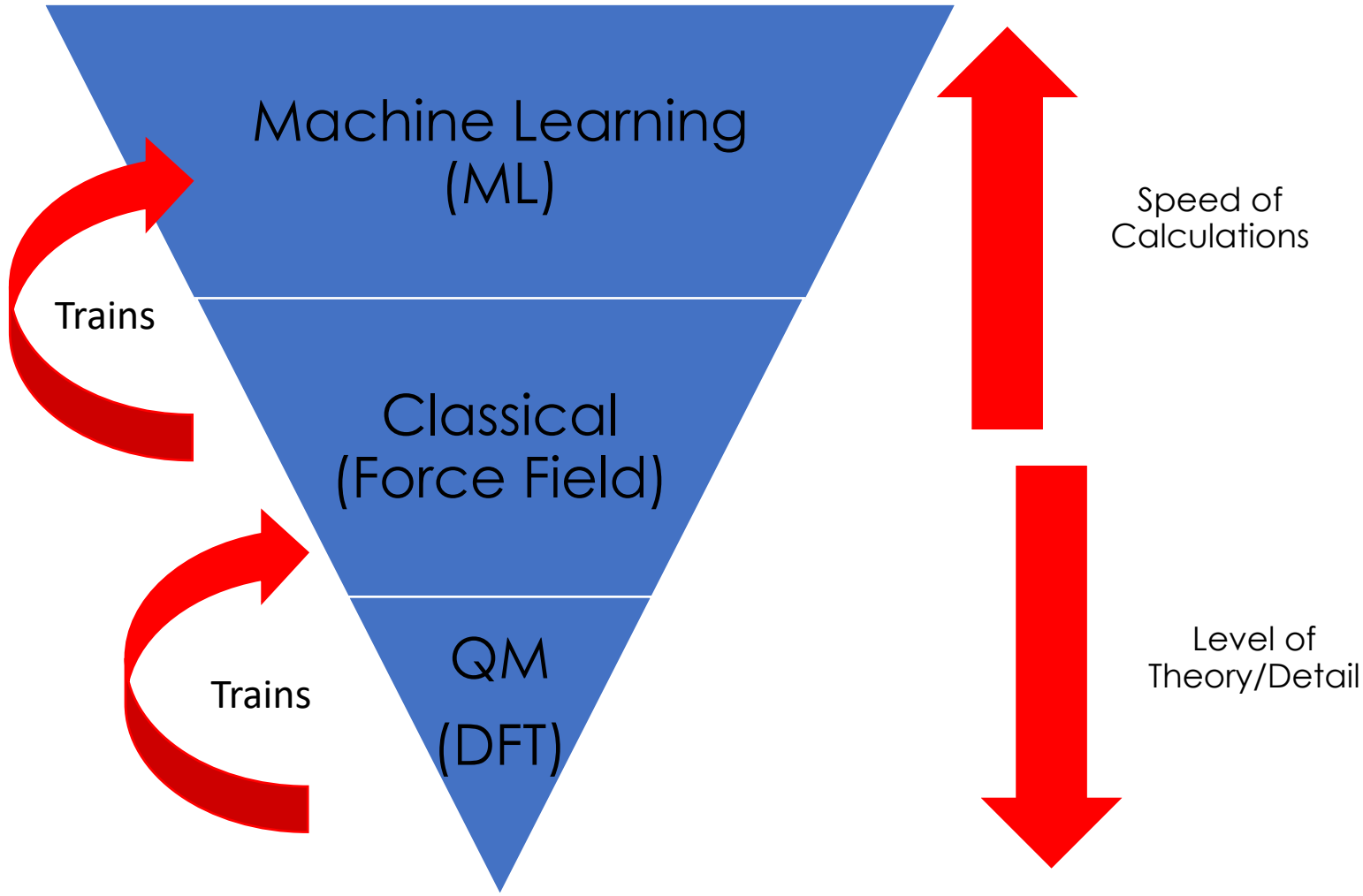
1. Construct a collection of sorbent structures and sorption conditions
2. Construct appropriate model potentials
3. Carry out computations to **estimate sorption** in a **representative subset** of the sorbent structures
4. Use AI/ML techniques to exploit relationships to **screen / design tailored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



Mordenite (MOR framework)

# Machine Learning for Materials Design

## Overall Strategy

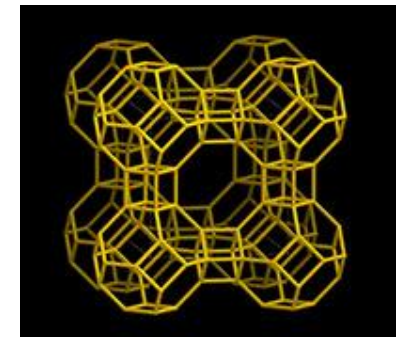
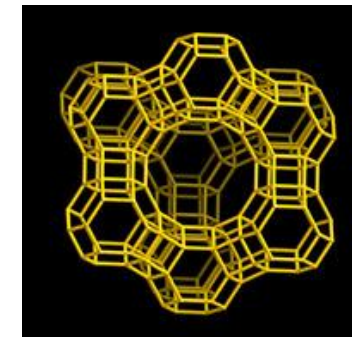


- **Machine Learning (ML):**
  - Accuracy depends on training set
  - Can be used for screening or design
  - Fastest of these three methods
- **Classical Simulations (FFs):**
  - Molecular simulations
  - Can study macroscopic properties
  - Useful for screening
- **Quantum Mechanical (QM) simulations:**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

## 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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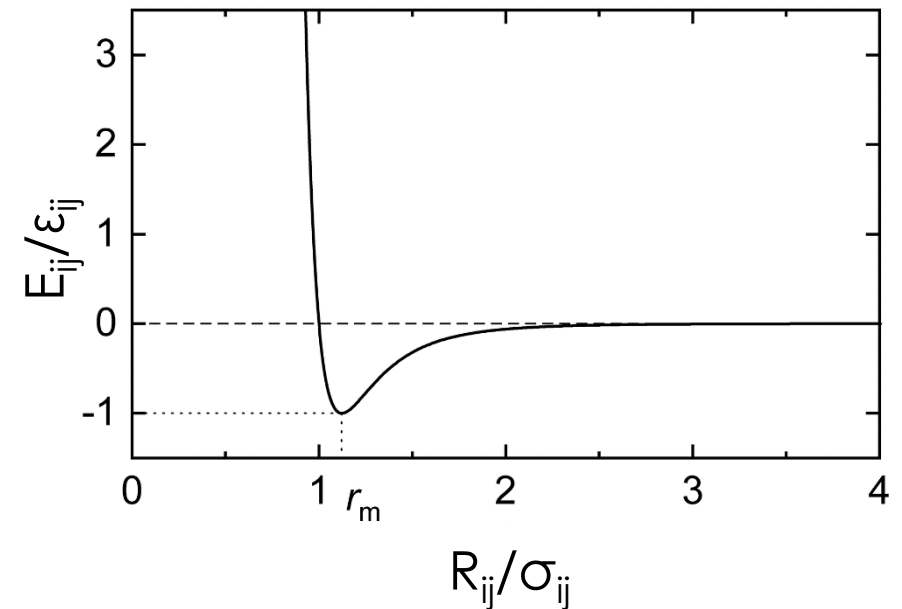




## 2. Construct Appropriate Model Potentials

- 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)
    - $\sigma$  is related to average atomic size
    - $\epsilon$  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)**

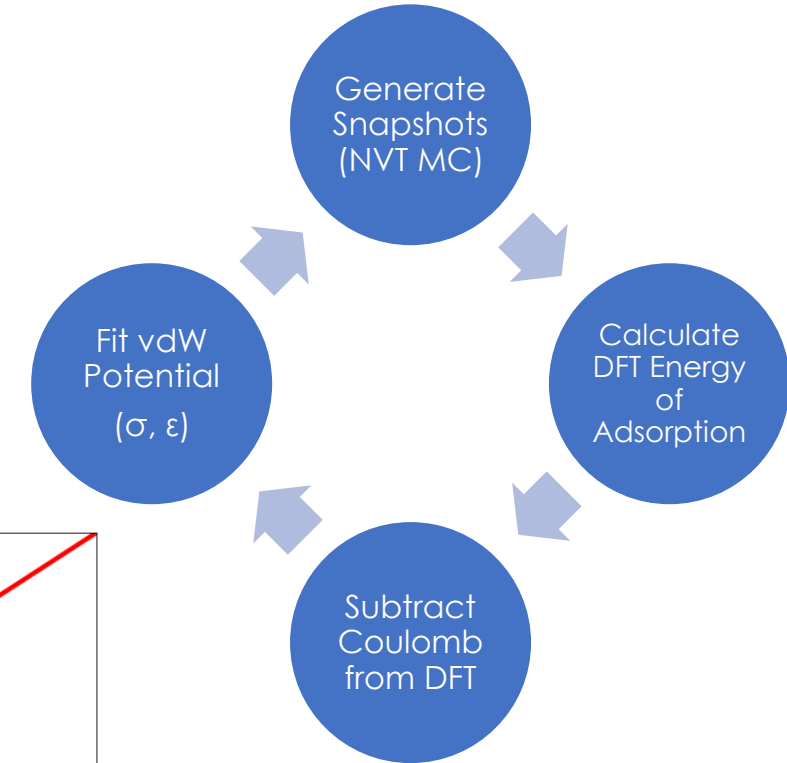
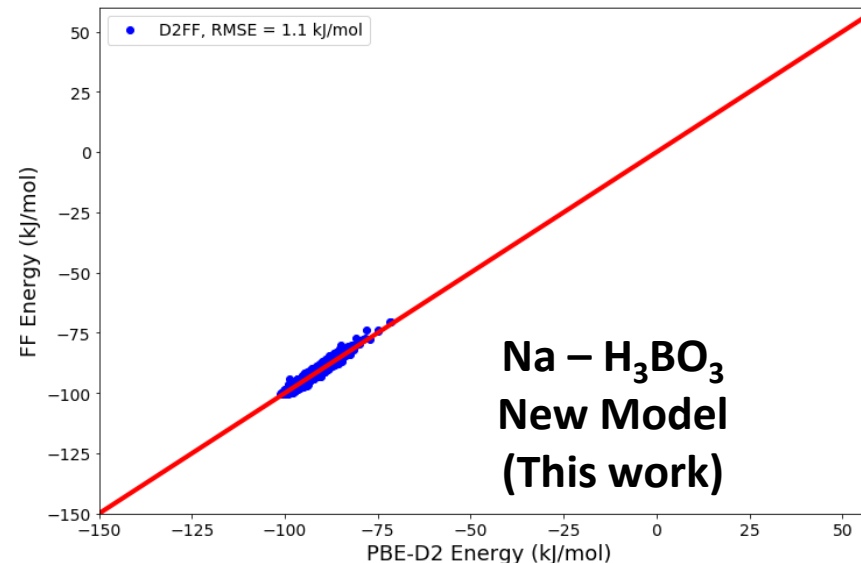
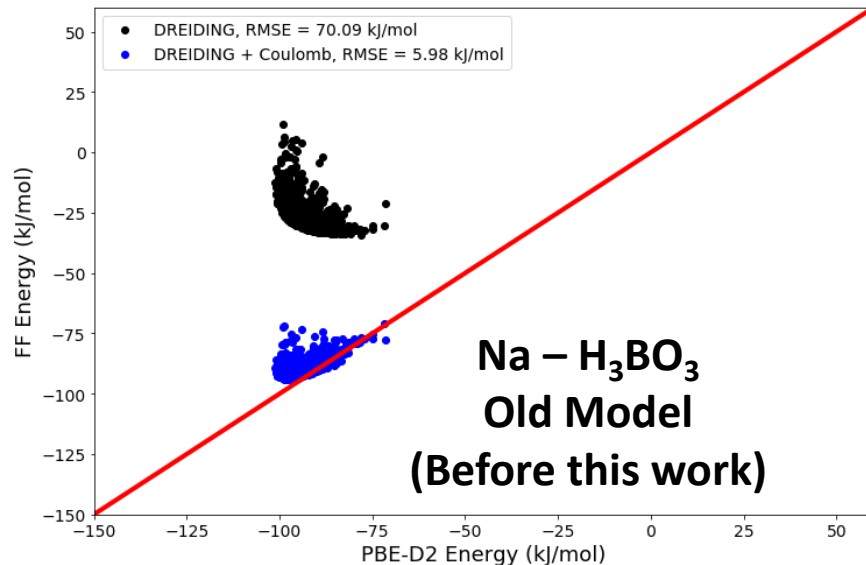
$$E_{ij}(R_{ij}) = 4\epsilon_{ij} \left[ \left( \frac{\sigma_{ij}}{R_{ij}} \right)^{12} - \left( \frac{\sigma_{ij}}{R_{ij}} \right)^6 \right]$$



# Fitting Adsorbate – Zeolite Interactions

## 2. Construct Appropriate Model Potentials

- DFT optimization showed **no chemical bonds broke** during adsorption of  $\text{H}_3\text{BO}_3$ ,  $\text{H}_2\text{O}$ 
  - Can use classical force fields for adsorption
- Complete for  $\text{Na}^+$ ,  $\text{K}^+$ ,  $\text{Ca}^{2+}$ ,  $\text{Mn}^{2+}$ ,  $\text{Fe}^{2+}$
- Construct models to reproduce adsorption energies from DFT

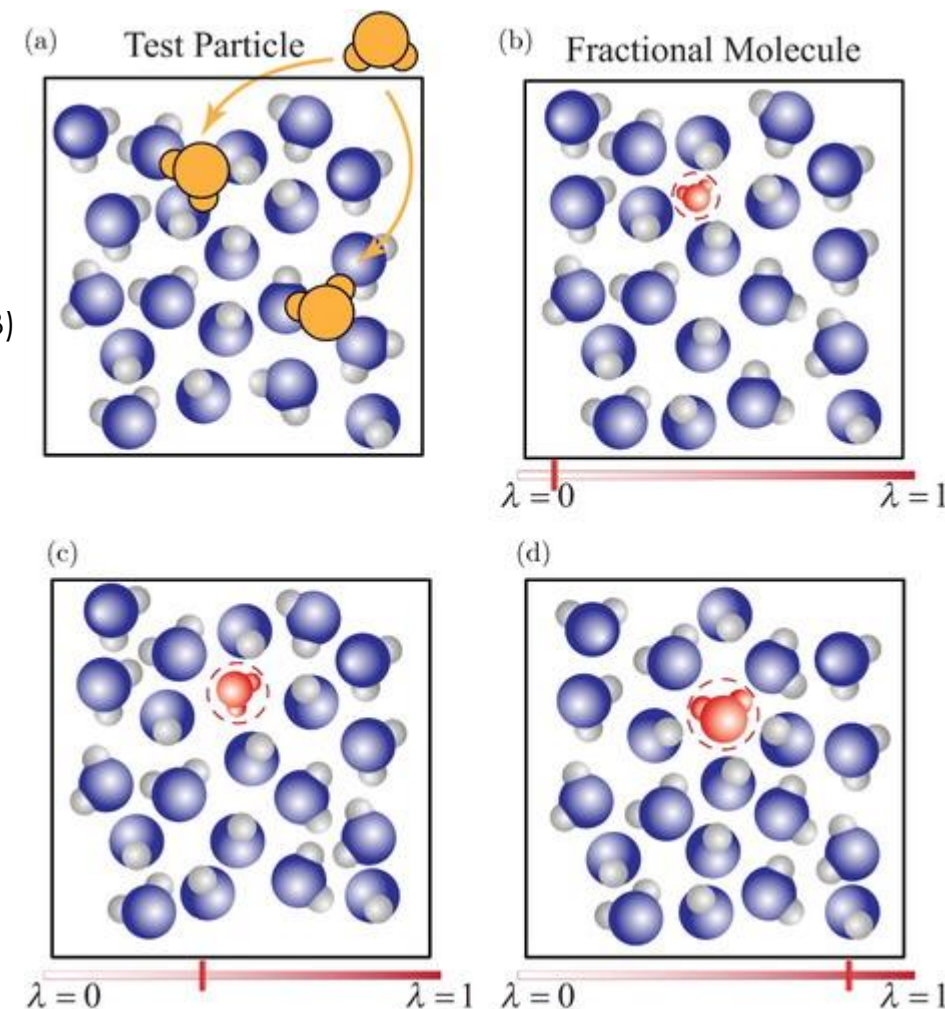


$$E_{ij}(R_{ij}) = 4\epsilon_{ij} \left[ \left( \frac{\sigma_{ij}}{R_{ij}} \right)^{12} - \left( \frac{\sigma_{ij}}{R_{ij}} \right)^6 \right] + \frac{q_i q_j}{R_{ij}}$$

# Adsorption Simulation Methods

## 3. Carry out Computations to Estimate Sorption in Sorbent Structures

- Adsorption from solution:
  - Reference state is now boric acid in solution
  - $f_A = \rho k_B T \left( e^{\mu_A^\infty / k_B T} \right) x_A$ 
    - $\mu_A^\infty$  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  $\lambda$  (0, 1]
  - Allows other molecules to move with fractional molecule present
  - Increases weight ( $\lambda$ ) until full molecule is grown



	$K_H$ (Pa*m <sup>3</sup> /mol)	Source
Experimental	$2.65 \cdot 10^{-7}$	HSDB (NIH)
<b>Sim (CFCMC-WI)</b>	<b><math>1.29 \cdot 10^{-7}</math></b>	<b>This work</b>

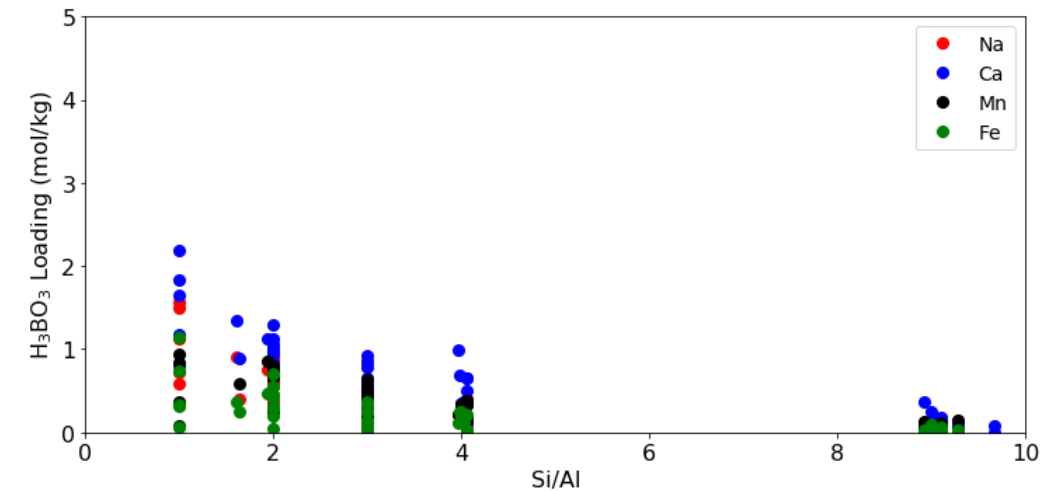
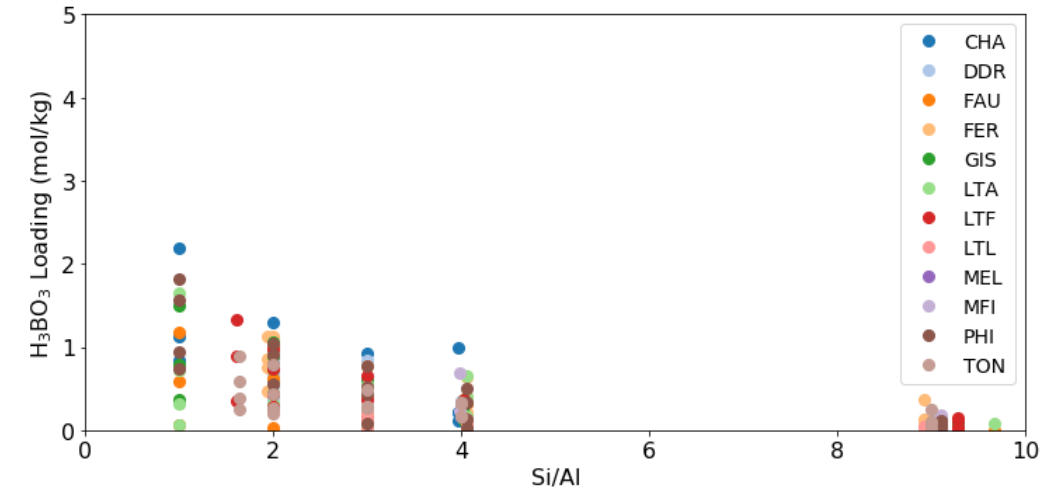
Rahbari, A., et al. "Recent advances in the continuous fractional component Monte Carlo methodology." *Molecular Simulation* 47.10-11 (2021): 804-823.

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

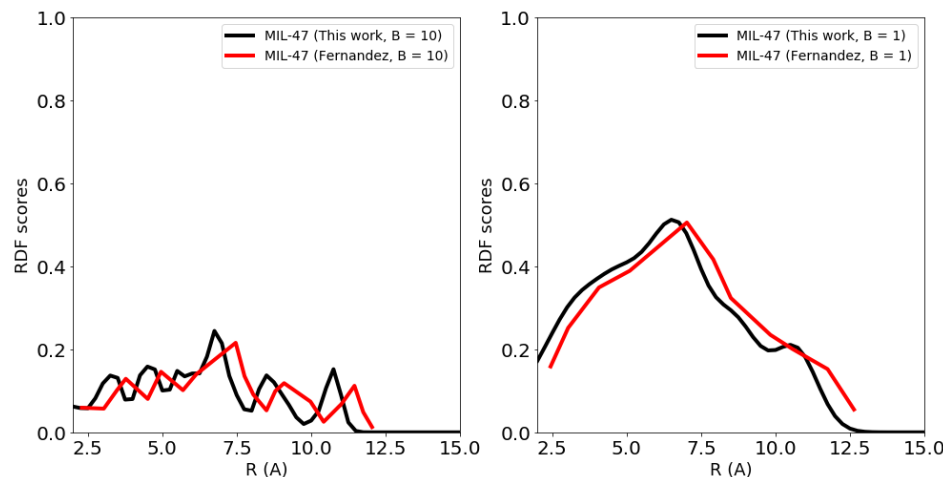
- Results for 1 ppm  $\text{H}_3\text{BO}_3$  solution
- Ca, Na-exchanged zeolites were top performers
  - Largest charge (Ca)
  - Most cations per unit cell (Na)
- Low Si/Al ratios performed best
  - Most cations per unit cell



## 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.



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

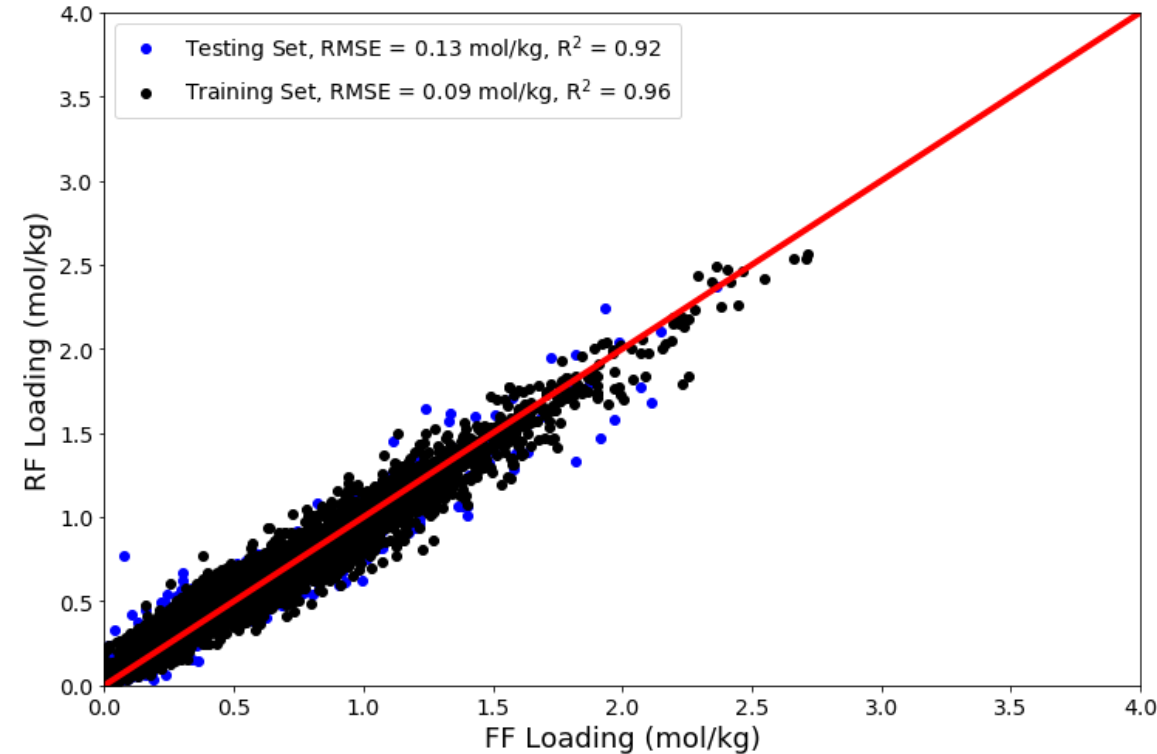
- ML Model requires:
  - **Descriptors:**
    - Crystal structure (topology)
    - Composition (both zeolite and solution)
  - **Adsorption data:**
    - Little experimental data available (not enough for ML)
    - Can predict adsorption using molecular simulations
      - Accuracy depends on models 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:**  $\text{H}_3\text{BO}_3$  uptake from 1-20 ppm  $\text{H}_3\text{BO}_3$  solution
- **Features:**
  - Stoichiometry (normalized by number of O atoms)
  - Six principal components for weighted RDFs (weighted by charge, electronegativity)
  - $\text{H}_3\text{BO}_3$  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  $\text{H}_3\text{BO}_3$  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

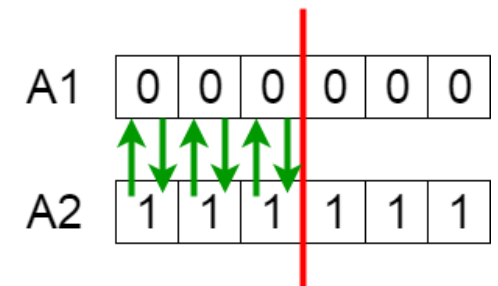
A5 

1	1	1	0	0	0
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After Mutation

A5 

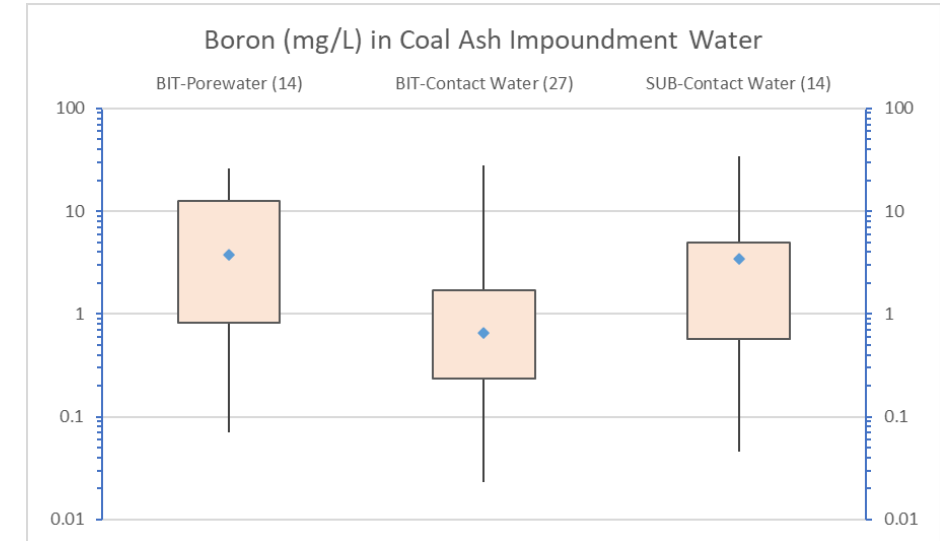
1	1	0	1	1	0
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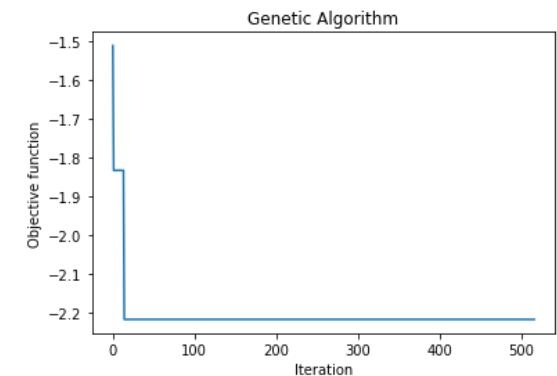
# Genetic Algorithms – Zeolite Optimization

## 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



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



# Questions?

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# NETL Resources

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CONTACT:

John Findley

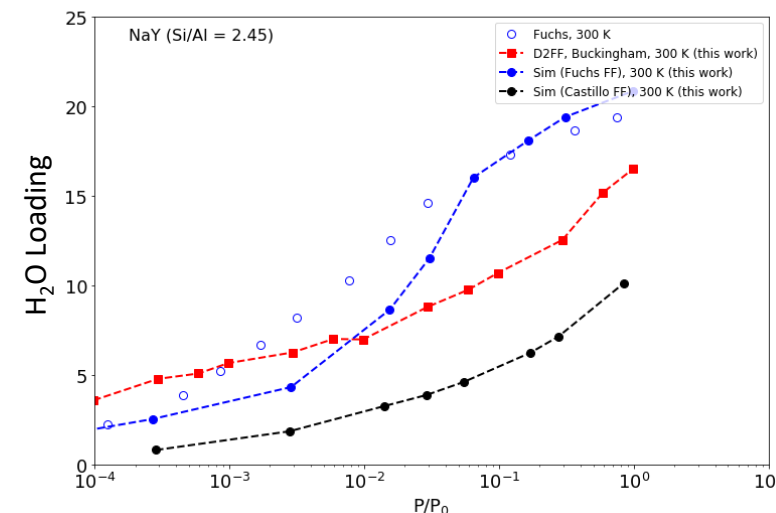
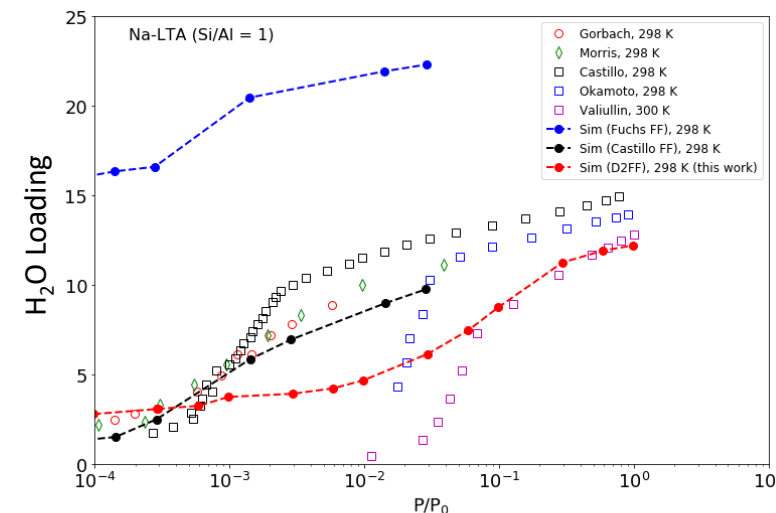
[John.Findley@netl.doe.gov](mailto:John.Findley@netl.doe.gov)



# Model Validation

## 2. Construct Appropriate Model Potentials (Validation)

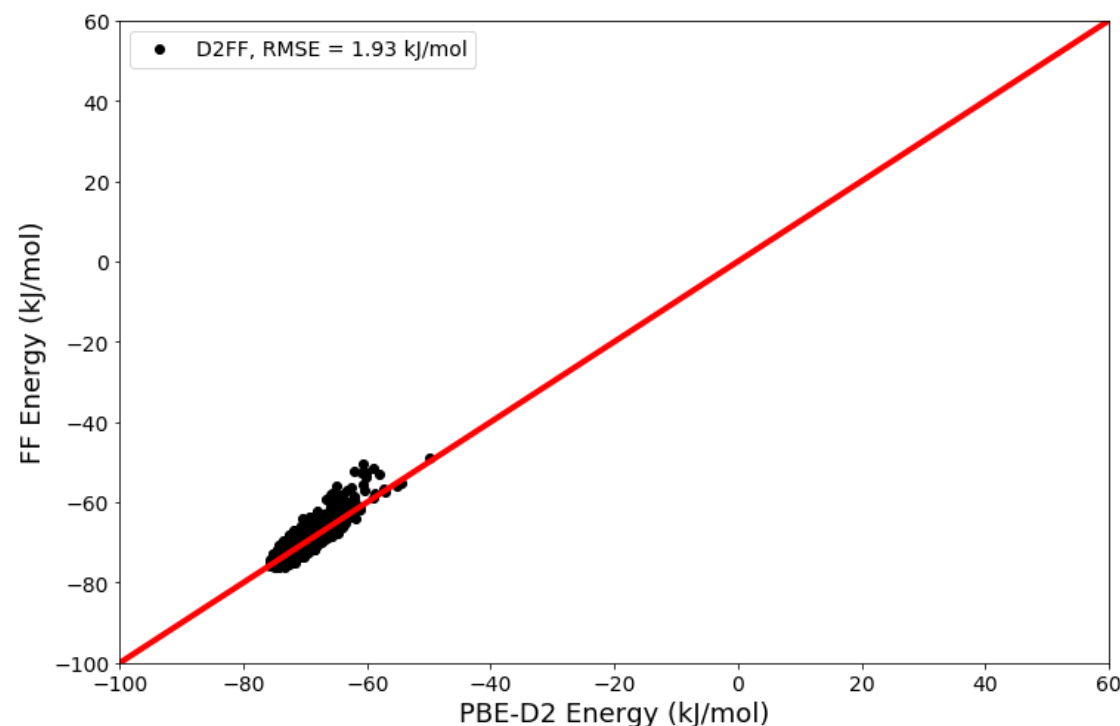
- Used models developed in December to predict water vapor adsorption isotherms
  - Model parametrized by fitting to Density Functional Theory Energies (PBE-D2 to be specific)
  - Compared two different potentials
    - Buckingham:**  $E_{ij} = A_{ij}e^{-B_{ij}R_{ij}} - \frac{C_{ij}}{R_{ij}^6} + \frac{q_i q_j}{4\pi\epsilon_0 R_{ij}}$
  - 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), H<sub>2</sub>O (finished), H<sub>3</sub>BO<sub>3</sub> (finished, pictured)



Atomic Pair	$A_{ij}$ (K)	$B_{ij}$ ( $\text{\AA}^{-1}$ )	$C_{ij}$ ( $\text{K}\cdot\text{\AA}^6$ )	$q_M$ (-e)	Source
K – Oz	$6.23\cdot 10^7$	3.43	$2.29\cdot 10^6$	0.990	Fang et al.
K – O (H <sub>2</sub> O)	$3.50\cdot 10^6$	2.55	0	0.990	This work

Atomic Pair	$\epsilon_{ij}$ (K)	$\sigma_{ij}$ ( $\text{\AA}$ )	$q_M$ (-e)
K – O (H <sub>3</sub> BO <sub>3</sub> )	348	2.87	0.990