Supplementary Material

Machine Learning for Prediction of CO₂/N₂/H₂O Selective **Adsorption and Separation in Metal-Zeolites**

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indicate if changes were made.



S1. Structural Model

The physical properties of CO₂, N₂ and H₂O are listed in Supplementary Table 1.

Physical properties	CO ₂	N ₂	H ₂ O
Kinetic diameter (Å)	3.3	3.6	2.7
Polarizability (× 10^{-25} cm ³)	26.3	17.7	14.8
Dipole moment (× 10 ¹⁸ esu cm)	0	0	1.87
Quadrupole moment (× 10^{-26} esy cm ²)	4.30	1.52	-

Supplementary Table 1. Physical properties of CO₂, N₂ and H₂O.

In this study, five kinds of topological structures (CHA, MFI, MOR, FER and FAU zeolites) were investigated for the CO₂ adsorption and separation. The supercell models for CHA ($2 \times 2 \times 2$), MOR ($1 \times 1 \times 2$), and FER ($1 \times 1 \times 2$) were used in calculation.^[1] To take the influence of the silicon-aluminum ratio (Si/Al) into consideration, models were established by replacing one/two/three/four silicon atoms with aluminum atoms, named as MFI-95, MFI-47, MFI-31, MFI-23, MOR-47, FER-35, CHA-47, and FAU-95, respectively.



Supplementary Figure 1. Structures of (a) CHA, (b) MFI, (c) MOR, (d) FER and (e) FAU zeolites.



Supplementary Figure 2. The proportion of five types of topological structures (MOR, MFI, CHA, FER, FAU) in the literatures.

The number of IZA structures has reached 264. There have been 197,763 reports related to zeolite from all databases on Web of science until August 14th, 2023. Moreover, the five kinds of topological structures have been studied widely according to literature researches (Supplementary Figure 2). Among them, researches about MOR take up 11.7%, and followed by MFI, accounting for 9.8%. CHA, FER and FAU have the proportion of 9.0%, 6.9% and 1.8%, respectively.



Supplementary Figure 3. The computational results of MFI-31 comparison between VASP and CP2K. In order to validate the computational results obtained by CP2K, MFI-31 zeolites were also performed using Vienna Ab initio Simulation Package (VASP), which gave similar results to CP2K.



Supplementary Figure 4. Pearson correlation coefficient matrix of descriptors for predicting (a) formation energy, (b) binding energy, (c) adsorption strength index.

S2. Machine Learning

Machine learning methods, ten algorithms, namely, least absolute shrinkage and selection operator (LASSO),^[2] support vector regression (SVR),^[3] gradient boosting regression (GBR), extreme gradient boosting regression (XGBoost), decisiontree,^[4] k-nearest neighbor (kNN),^[5] linear ridge,^[6] multiple Linear regression (MLR),^[7] artificial neural network (ANN),^[8] and ExtraTrees^[9], were chosen to compare the performance of the prediction models. The LASSO is a penalized likelihood approach. The SVR algorithm applies the squared ϵ -insensitive loss function, which makes the optimization problem strictly convex and obtains a clear solution. The GBR and XGBoost are ensemble boosting learning models. ^[10] The ensemble learning algorithm can train multiple constituent learning algorithms at onetime to obtain better prediction compared with a single-constituent learner. Three evaluation indexes, mean absolute error (MAE), root mean square error (RMSE), and coefficient of determination (R²) are used to evaluate the predictive capacity of ML models. ML trainings were implemented by the aid of scikit-learn package. ^[11] ExtreTrees, GBR and XGBoost showed the good performance on the formation energy, binding energy and adsorption strength index.



Supplementary Figure 5. Comparison of the performance with 10 algorithms of the FL in (a) formation energy, (b) binding energy, (c) adsorption strength index.



Supplementary Figure 6. The calculated binding energies of adsorbates on the metal-zeolites (a) MFI-95, (b) MFI-47, (c) MFI-31.



Supplementary Figure 7. The calculated binding energies of adsorbates on the metal-zeolites (a) MFI-23, (b) MOR-47, (c) FER-35.



Supplementary Figure 8. The calculated binding energies of adsorbates on the metal-zeolites (a) CHA-47, (b) FAU-95.

The binding energies of 5 kinds of zeolites with embedded metals were calculated to study the adsorption strength of adsorbates. The embedded metal atoms are involved in 3*d*, 4*d*, and 5*d* transition metal. As shown in Supplementary Figures 6-8, the Mo, Zr-zeolite preferred to adsorb the CO_2 in comparison of the N₂ and H₂O. Zeolites with 4*d* transition metal embedded showed stronger binding ability than those of 3*d* and 5*d* transition metals.

S3. Feature Selection and the Dataset

To find out the good materials to separate the CO₂ from the mixed gases (N₂/H₂O) with the similar kinetic diameters, there are 15 features chosen to build the machine learning models to predict the formation energy, binding energy and adsorption strength index. To describe the influence of metal atoms, the metal atomic electronegativity (χ), first ionization energy of metal atoms (IE), the number of *d* electrons (Nd) and metal atomic radius (R) were chosen. The more electronegative the metal is, the easier the metal-zeolite are of attracting electron. The first ionization energy of metal atom is the energy required for a gaseous atom in its ground state to lose an electron in its outermost shell. It is easier to lose an electron when the atom has a lower first ionization. Nd is the number of *d* orbitals electrons of metal atoms in the valence shell. The refinement distance least squares (R_{DLS}) was considered to describe the geometry distortion of zeolite framework. The geometry distortion may play a part in adsorption process. Furthermore, changing the Si/Al ratio of the zeolites can also tune the electrostatic environment. The adsorption and diffusion properties are related to the channel architecture of zeolites. The pore limiting diameter (PLD) of zeolite was applied to describe the porosity of zeolites. All of the features are listed in the Supplementary Table 2 and the data are listed in the Supplementary Table 3.

abbreviation	name	unit
PLD	Pore limiting diameter of zeolite	Å
Si/Al	Number ratio of silicon to aluminum	_
R _{DLS}	Refinement distance least-squares	_
Polar	polarizability of adsorbate	cm ³
Np	Valence electron number of the adsorbate	_
d	Kinetic diameter of the adsorbate	Å
NG	The number of group in the periodic table of metal	_
χ	Electronegativity	_
IE	The first ionization energy of metal atom	kJ mol ⁻¹
Nd	d shell electron number of metal atom	_
AN	Atomic number of metal	_
R	Metal atomic radius	pm
$\Delta_{fus}H$	Enthalpy of fusion of metal	kJ mol ⁻¹
PLD/d	Ratio of pore limiting diameter of zeolite to kinetic diameter of adsorbate	_

Supplementary Table 2. List of features used in prediction of the formation energy, binding energy and adsorption strength index.

System	Si/Al	PLD/d	PLD (Å)	NG	Polar ($\times 10^{-25}$ cm ³)	Nd	Np	$\Delta_{\rm fus} H$ (kJ mol ⁻¹)
CO ₂ @Sc-MFI-95	95	1.927273	6.36	3	26.3	1	16	16.11
CO ₂ @Ti-MFI-95	95	1.927273	6.36	4	26.3	2	16	18.6
CO ₂ @V-MFI-95	95	1.927273	6.36	5	26.3	3	16	20.8
CO2@Cr-MFI-95	95	1.927273	6.36	6	26.3	5	16	20
CO2@Mn-MFI-95	95	1.927273	6.36	7	26.3	5	16	14.64
CO2@Fe-MFI-95	95	1.927273	6.36	8	26.3	6	16	13.8
CO2@Co-MFI-95	95	1.927273	6.36	9	26.3	7	16	16.19
CO2@Ni-MFI-95	95	1.927273	6.36	10	26.3	8	16	17.2
CO2@Cu-MFI-95	95	1.927273	6.36	11	26.3	10	16	13.14
CO ₂ @Zn-MFI-95	95	1.927273	6.36	12	26.3	10	16	7.38
CO ₂ @Y-MFI-95	95	1.927273	6.36	3	26.3	1	16	17.5
CO2@Zr-MFI-95	95	1.927273	6.36	4	26.3	2	16	21
CO ₂ @Nb-MFI-95	95	1.927273	6.36	5	26.3	4	16	26.9
CO2@Mo-MFI-95	95	1.927273	6.36	6	26.3	5	16	36
CO2@Ru-MFI-95	95	1.927273	6.36	8	26.3	7	16	25.5
CO2@Rh-MFI-95	95	1.927273	6.36	9	26.3	8	16	21.8
CO2@Pd-MFI-95	95	1.927273	6.36	10	26.3	10	16	16.74
CO ₂ @Ag-MFI-95	95	1.927273	6.36	11	26.3	10	16	11.3
CO2@Cd-MFI-95	95	1.927273	6.36	12	26.3	10	16	6.07
CO2@Ta-MFI-95	95	1.927273	6.36	5	26.3	3	16	36
CO ₂ @W-MFI-95	95	1.927273	6.36	6	26.3	4	16	35

Supplementary Table 3. Part of data (20%) in dataset of machine learning to predict the binding energy and adsorption strength index.

C (S:/A1		PLD	N	Polar	NT 1	N	$\Delta_{ m fus} H$
System	S1/AI	PLD/d	(Å)	NG	$(\times 10^{-25} \text{ cm}^3)$	Nd	Np	(kJ mol ⁻¹)
CO ₂ @Os-MFI-95	95	1.927273	6.36	8	26.3	6	16	29.29
CO ₂ @Ir-MFI-95	95	1.927273	6.36	9	26.3	7	16	26.36
CO2@Pt-MFI-95	95	1.927273	6.36	10	26.3	9	16	19.66
CO ₂ @Pb-MFI-95	95	1.927273	6.36	14	26.3	0	16	4.77
N ₂ @Sc-MFI-95	95	1.766667	6.36	3	17.7	1	10	16.11
N ₂ @Ti-MFI-95	95	1.766667	6.36	4	17.7	2	10	18.6
$N_2@V-MFI-95$	95	1.766667	6.36	5	17.7	3	10	20.8
N2@Cr-MFI-95	95	1.766667	6.36	6	17.7	5	10	20
N2@Mn-MFI-95	95	1.766667	6.36	7	17.7	5	10	14.64
N ₂ @Fe-MFI-95	95	1.766667	6.36	8	17.7	6	10	13.8
N ₂ @Co-MFI-95	95	1.766667	6.36	9	17.7	7	10	16.19
N2@Ni-MFI-95	95	1.766667	6.36	10	17.7	8	10	17.2
N ₂ @Cu-MFI-95	95	1.766667	6.36	11	17.7	10	10	13.14
N2@Zn-MFI-95	95	1.766667	6.36	12	17.7	10	10	7.38
$N_2@Y-MFI-95$	95	1.766667	6.36	3	17.7	1	10	17.5
N ₂ @Zr-MFI-95	95	1.766667	6.36	4	17.7	2	10	21
N ₂ @Nb-MFI-95	95	1.766667	6.36	5	17.7	4	10	26.9
N ₂ @Mo-MFI-95	95	1.766667	6.36	6	17.7	5	10	36
N ₂ @Ru-MFI-95	95	1.766667	6.36	8	17.7	7	10	25.5
N2@Rh-MFI-95	95	1.766667	6.36	9	17.7	8	10	21.8
N ₂ @Pd-MFI-95	95	1.766667	6.36	10	17.7	10	10	16.74

(Supplementary Table 3. Continued)

Sautan	S:/A1		PLD	N	Polar	NT 1	Nn	$\Delta_{ m fus} H$
System	51/AI	PLD/0	(Å)	ING	$(\times 10^{-25} \text{cm}^3)$	ING	мр	(kJ mol ⁻¹)
N ₂ @Cd-MFI-95	95	1.766667	6.36	12	17.7	10	10	6.07
N ₂ @Ta-MFI1-95	95	1.766667	6.36	5	17.7	3	10	36
N ₂ @W-MFI-95	95	1.766667	6.36	6	17.7	4	10	35
N ₂ @Re-MFI-95	95	1.766667	6.36	7	17.7	5	10	33.05
N ₂ @Os-MFI-95	95	1.766667	6.36	8	17.7	6	10	29.29
N ₂ @Ir-MFI-95	95	1.766667	6.36	9	17.7	7	10	26.36
N ₂ @Pt-MFI-95	95	1.766667	6.36	10	17.7	9	10	19.66
N ₂ @Pb-MFI-95	95	1.766667	6.36	14	17.7	0	10	4.77
H ₂ O@Sc-MFI-95	95	2.355556	6.36	3	14.8	1	8	16.11
H ₂ O@Ti-MFI-95	95	2.355556	6.36	4	14.8	2	8	18.6
H ₂ O@V-MFI-95	95	2.355556	6.36	5	14.8	3	8	20.8
H ₂ O@Cr-MFI-95	95	2.355556	6.36	6	14.8	5	8	20
H ₂ O@Mn-MFI-95	95	2.355556	6.36	7	14.8	5	8	14.64
H ₂ O@Fe-MFI-95	95	2.355556	6.36	8	14.8	6	8	13.8
H ₂ O@Co-MFI-95	95	2.355556	6.36	9	14.8	7	8	16.19
H ₂ O@Ni-MFI-95	95	2.355556	6.36	10	14.8	8	8	17.2
H ₂ O@Cu-MFI-95	95	2.355556	6.36	11	14.8	10	8	13.14
H ₂ O@Zn-MFI-95	95	2.355556	6.36	12	14.8	10	8	7.38
H ₂ O@Y-MFI-95	95	2.355556	6.36	3	14.8	1	8	17.5
H ₂ O@Zr-MFI-95	95	2.355556	6.36	4	14.8	2	8	21
H ₂ O@Nb-MFI-95	95	2.355556	6.36	5	14.8	4	8	26.9

(Supplementary Table 3. Continued)

Constant	S ;/A1		PLD	N	Polar	NLI	Nn	$\Delta_{ m fus} H$
System	51/A1	PLD/d	(Å)	ING	$(\times 10^{-25} \text{ cm}^3)$	INd	мр	(kJ mol ⁻¹)
H ₂ O@Ru-MFI-95	95	2.355556	6.36	8	14.8	7	8	25.5
H ₂ O@Rh-MFI-95	95	2.355556	6.36	9	14.8	8	8	21.8
H ₂ O@Pd-MFI-95	95	2.355556	6.36	10	14.8	10	8	16.74
H ₂ O@Ag-MFI-95	95	2.355556	6.36	11	14.8	10	8	11.3
H ₂ O@Cd-MFI-95	95	2.355556	6.36	12	14.8	10	8	6.07
H ₂ O@Ta-MFI-95	95	2.355556	6.36	5	14.8	3	8	36
H ₂ O@W-MFI-95	95	2.355556	6.36	6	14.8	4	8	35
H ₂ O@Re-MFI-95	95	2.355556	6.36	7	14.8	5	8	33.05
H ₂ O@Os-MFI-95	95	2.355556	6.36	8	14.8	6	8	29.29
H ₂ O@Ir-MFI-95	95	2.355556	6.36	9	14.8	7	8	26.36
H ₂ O@Pt-MFI-95	95	2.355556	6.36	10	14.8	9	8	19.66
H ₂ O@Pb-MFI-95	95	2.355556	6.36	14	14.8	0	8	4.77
CO ₂ @Sc-MFI-47	47	1.927273	6.36	3	26.3	1	16	16.11
CO ₂ @Ti-MFI-47	47	1.927273	6.36	4	26.3	2	16	18.6
CO ₂ @V-MFI-47	47	1.927273	6.36	5	26.3	3	16	20.8
CO ₂ @Cr-MFI-47	47	1.927273	6.36	6	26.3	5	16	20
CO2@Mn-MFI-47	47	1.927273	6.36	7	26.3	5	16	14.64
CO ₂ @Fe-MFI-47	47	1.927273	6.36	8	26.3	6	16	13.8
CO ₂ @Co-MFI-47	47	1.927273	6.36	9	26.3	7	16	16.19
CO ₂ @Ni-MFI-47	47	1.927273	6.36	10	26.3	8	16	17.2
CO ₂ @Cu-MFI-47	47	1.927273	6.36	11	26.3	10	16	13.14

(Supplementary Table 3. Continued)

Sautan	S:/A1		PLD	N	Polar	N J	Nn	$\Delta_{ m fus} H$
System	51/A1	PLD/0	(Å)	ING	$(\times 10^{-25} \text{cm}^3)$	ING	мр	(kJ mol ⁻¹)
CO ₂ @Y-MFI-47	47	1.927273	6.36	3	26.3	1	16	17.5
CO ₂ @Zr-MFI-47	47	1.927273	6.36	4	26.3	2	16	21
CO ₂ @Nb-MFI-47	47	1.927273	6.36	5	26.3	4	16	26.9
CO2@Mo-MFI-47	47	1.927273	6.36	6	26.3	5	16	36
CO ₂ @Ru-MFI-47	47	1.927273	6.36	8	26.3	7	16	25.5
CO2@Rh-MFI-47	47	1.927273	6.36	9	26.3	8	16	21.8
CO ₂ @Pd-MFI-47	47	1.927273	6.36	10	26.3	10	16	16.74
CO ₂ @Ag-MFI-47	47	1.927273	6.36	11	26.3	10	16	11.3
CO2@Cd-MFI-47	47	1.927273	6.36	12	26.3	10	16	6.07
CO ₂ @Ta-MFI-47	47	1.927273	6.36	5	26.3	3	16	36
CO ₂ @W-MFI-47	47	1.927273	6.36	6	26.3	4	16	35
CO ₂ @Re-MFI-47	47	1.927273	6.36	7	26.3	5	16	33.05
CO ₂ @Os-MFI-47	47	1.927273	6.36	8	26.3	6	16	29.29
CO ₂ @Ir-MFI-47	47	1.927273	6.36	9	26.3	7	16	26.36
CO ₂ @Pt-MFI-47	47	1.927273	6.36	10	26.3	9	16	19.66
CO ₂ @Pb-MFI-47	47	1.927273	6.36	14	26.3	0	16	4.77
CO2@Sc-MFI-47	47	1.766667	6.36	3	17.7	1	10	16.11
N ₂ @Ti-MFI-47	47	1.766667	6.36	4	17.7	2	10	18.6
N ₂ @V-MFI-47	47	1.766667	6.36	5	17.7	3	10	20.8
N ₂ @Cr-MFI-47	47	1.766667	6.36	6	17.7	5	10	20
N ₂ @Mn-MFI-47	47	1.766667	6.36	7	17.7	5	10	14.64

(Supplementary Table 3. Continued)

System	d	R _{DL} a	ΔN	24	R	IE	$E_{ m b}$	I
System	(Å)	NDLS	AIN	λ	(pm)	(kJ mol ⁻¹)	(eV)	1
CO ₂ @Sc-MFI-95	3.3	0.002	21	1.36	144	633.1	-0.76	0.43
CO ₂ @Ti-MFI-95	3.3	0.002	22	1.54	132	658.8	-0.78	0.35
CO ₂ @V-MFI-95	3.3	0.002	23	1.63	122	650.9	-1.28	0.33
CO ₂ @Cr-MFI-95	3.3	0.002	24	1.66	118	652.9	-0.25	0.18
CO ₂ @Mn-MFI-95	3.3	0.002	25	1.55	117	717.3	-1.13	0.29
CO ₂ @Fe-MFI-95	3.3	0.002	26	1.83	117	762.5	-0.76	0.22
CO ₂ @Co-MFI-95	3.3	0.002	27	1.88	116	760.4	-0.47	0.19
CO ₂ @Ni-MFI-95	3.3	0.002	28	1.91	115	737.1	-0.45	0.24
CO ₂ @Cu-MFI-95	3.3	0.002	29	1.9	117	745.5	-0.33	0.20
CO ₂ @Zn-MFI-95	3.3	0.002	30	1.65	125	906.4	-0.38	0.25
CO ₂ @Y-MFI-95	3.3	0.002	39	1.22	162	600	-0.98	0.37
CO ₂ @Zr-MFI-95	3.3	0.002	40	1.33	145	640.1	-0.57	0.39
CO ₂ @Nb-MFI-95	3.3	0.002	41	1.6	134	652.1	-1.41	0.58
CO2@Mo-MFI-95	3.3	0.002	42	2.16	130	684.3	-1.53	0.36
CO ₂ @Ru-MFI-95	3.3	0.002	44	2.2	125	710.2	-0.30	0.20
CO2@Rh-MFI-95	3.3	0.002	45	2.28	125	719.7	-0.08	0.18
CO ₂ @Pd-MFI-95	3.3	0.002	46	2.2	128	804.4	-0.33	0.19
CO ₂ @Ag-MFI-95	3.3	0.002	47	1.93	134	731	-0.56	0.24
CO ₂ @Cd-MFI-95	3.3	0.002	48	1.69	148	867.8	-0.35	0.23
CO ₂ @Ta-MFI-95	3.3	0.002	73	1.5	134	761	-2.09	0.53
CO ₂ @W-MFI-95	3.3	0.002	74	2.36	130	770	-1.63	0.32

(Supplementary Table 3. Continued)

System	d	Para	ΔN	24	R	IE	E_{b}	I
System	(Å)	RDLS	AI	λ	(pm)	(kJ mol ⁻¹)	(eV)	1
CO ₂ @Os-MFI-95	3.3	0.002	76	2.2	126	840	-0.69	0.30
CO ₂ @Ir-MFI-95	3.3	0.002	77	2.2	127	880	-0.32	0.24
CO ₂ @Pt-MFI-95	3.3	0.002	78	2.28	130	870	-0.32	0.16
CO ₂ @Pb-MFI-95	3.3	0.002	82	1.8	147	715.6	-0.45	0.25
N ₂ @Sc-MFI-95	3.6	0.002	21	1.36	144	633.1	0.03	0.19
N ₂ @Ti-MFI-95	3.6	0.002	22	1.54	132	658.8	-0.94	0.42
N ₂ @V-MFI-95	3.6	0.002	23	1.63	122	650.9	-1.70	0.50
N ₂ @Cr-MFI-95	3.6	0.002	24	1.66	118	652.9	-1.41	0.56
N ₂ @Mn-MFI-95	3.6	0.002	25	1.55	117	717.3	-1.62	0.48
N ₂ @Fe-MFI-95	3.6	0.002	26	1.83	117	762.5	-1.65	0.54
N ₂ @Co-MFI-95	3.6	0.002	27	1.88	116	760.4	-1.52	0.54
N ₂ @Ni-MFI-95	3.6	0.002	28	1.91	115	737.1	-1.15	0.48
N ₂ @Cu-MFI-95	3.6	0.002	29	1.9	117	745.5	-1.07	0.42
N ₂ @Zn-MFI-95	3.6	0.002	30	1.65	125	906.4	-0.37	0.25
N ₂ @Y-MFI-95	3.6	0.002	39	1.22	162	600	0.58	0.08
N ₂ @Zr-MFI-95	3.6	0.002	40	1.33	145	640.1	-0.60	0.40
N ₂ @Nb-MFI-95	3.6	0.002	41	1.6	134	652.1	-0.19	0.17
N ₂ @Mo-MFI-95	3.6	0.002	42	2.16	130	684.3	-1.77	0.46
N ₂ @Ru-MFI-95	3.6	0.002	44	2.2	125	710.2	-1.31	0.54
N ₂ @Rh-MFI-95	3.6	0.002	45	2.28	125	719.7	-1.02	0.46
N ₂ @Pd-MFI-95	3.6	0.002	46	2.2	128	804.4	-0.98	0.36

(Supplementary Table 3. Continued)

System	d	P	AN		R	IE	E_{b}	I
System	(Å)	K _{DLS}	AN	χ	(pm)	(kJ mol ⁻¹)	(eV)	1
N ₂ @Cd-MFI-95	3.6	0.002	48	1.69	148	867.8	-0.49	0.27
N ₂ @Ta-MFI-95	3.6	0.002	73	1.5	134	761	-1.67	0.35
N ₂ @W-MFI-95	3.6	0.002	74	2.36	130	770	-2.25	0.59
N ₂ @Re-MFI-95	3.6	0.002	75	1.9	128	760	-1.65	0.43
N ₂ @Os-MFI-95	3.6	0.002	76	2.2	126	840	-1.30	0.55
N ₂ @Ir-MFI-95	3.6	0.002	77	2.2	127	880	-1.12	0.53
N ₂ @Pt-MFI-95	3.6	0.002	78	2.28	130	870	-1.49	0.50
N ₂ @Pb-MFI-95	3.6	0.002	82	1.8	147	715.6	-0.49	0.26
H ₂ O@Sc-MFI-95	2.7	0.002	21	1.36	144	633.1	-0.63	0.38
H ₂ O@Ti-MFI-95	2.7	0.002	22	1.54	132	658.8	-0.34	0.23
H ₂ O@V-MFI-95	2.7	0.002	23	1.63	122	650.9	-0.66	0.18
H ₂ O@Cr-MFI-95	2.7	0.002	24	1.66	118	652.9	-0.67	0.27
H ₂ O@Mn-MFI-95	2.7	0.002	25	1.55	117	717.3	-0.85	0.22
H ₂ O@Fe-MFI-95	2.7	0.002	26	1.83	117	762.5	-0.81	0.23
H ₂ O@Co-MFI-95	2.7	0.002	27	1.88	116	760.4	-0.85	0.27
H ₂ O@Ni-MFI-95	2.7	0.002	28	1.91	115	737.1	-0.59	0.28
H ₂ O@Cu-MFI-95	2.7	0.002	29	1.9	117	745.5	-0.98	0.38
H ₂ O@Zn-MFI-95	2.7	0.002	30	1.65	125	906.4	-1.09	0.51
H ₂ O@Y-MFI-95	2.7	0.002	39	1.22	162	600	-1.39	0.55
H ₂ O@Zr-MFI-95	2.7	0.002	40	1.33	145	640.1	0.04	0.21
H ₂ O@Nb-MFI-95	2.7	0.002	41	1.6	134	652.1	-0.56	0.25

(Supplementary Table 3. Continued)

Sustam	d	D	AN		R	IE	$E_{\rm b}$	T
System	(Å)	K _{DLS}	AIN	χ	(pm)	(kJ mol ⁻¹)	(eV)	1
H ₂ O@Ru-MFI-95	2.7	0.002	44	2.2	125	710.2	-0.59	0.26
H ₂ O@Rh-MFI-95	2.7	0.002	45	2.28	125	719.7	-0.78	0.36
H ₂ O@Pd-MFI-95	2.7	0.002	46	2.2	128	804.4	-1.19	0.45
H ₂ O@Ag-MFI-95	2.7	0.002	47	1.93	134	731	-1.17	0.44
H ₂ O@Cd-MFI-95	2.7	0.002	48	1.69	148	867.8	-1.11	0.50
H ₂ O@Ta-MFI-95	2.7	0.002	73	1.5	134	761	-0.61	0.12
H ₂ O@W-MFI-95	2.7	0.002	74	2.36	130	770	-0.38	0.09
H ₂ O@Re-MFI-95	2.7	0.002	75	1.9	128	760	-0.30	0.11
H ₂ O@Os-MFI-95	2.7	0.002	76	2.2	126	840	0.03	0.15
H ₂ O@Ir-MFI-95	2.7	0.002	77	2.2	127	880	-0.29	0.23
H ₂ O@Pt-MFI-95	2.7	0.002	78	2.28	130	870	-1.11	0.34
H ₂ O@Pb-MFI-95	2.7	0.002	82	1.8	147	715.6	-1.15	0.50
CO ₂ @Sc-MFI-47	3.3	0.002	21	1.36	144	633.1	-0.93	0.28
CO ₂ @Ti-MFI-47	3.3	0.002	22	1.54	132	658.8	-1.06	0.28
CO ₂ @V-MFI-47	3.3	0.002	23	1.63	122	650.9	-0.78	0.23
CO ₂ @Cr-MFI-47	3.3	0.002	24	1.66	118	652.9	-0.03	0.16
CO ₂ @Mn-MFI-47	3.3	0.002	25	1.55	117	717.3	-0.12	0.14
CO ₂ @Fe-MFI-47	3.3	0.002	26	1.83	117	762.5	-0.36	0.20
CO ₂ @Co-MFI-47	3.3	0.002	27	1.88	116	760.4	-0.39	0.21
CO ₂ @Ni-MFI-47	3.3	0.002	28	1.91	115	737.1	-0.29	0.20
CO ₂ @Cu-MFI-47	3.3	0.002	29	1.9	117	745.5	-0.41	0.22

(Supplementary Table 3. Continued)

	d	D	4 N T			IE	E_{b}		
System	(Å)	K _{DLS}	AN	χ	(pm)	(kJ mol ⁻¹)	(eV)	Ι	
CO ₂ @Y-MFI-47	3.3	0.002	39	1.22	162	600	-1.32	0.36	
CO ₂ @Zr-MFI-47	3.3	0.002	40	1.33	145	640.1	-2.35	0.57	
CO ₂ @Nb-MFI-47	3.3	0.002	41	1.6	134	652.1	-2.15	0.49	
CO ₂ @Mo-MFI-47	3.3	0.002	42	2.16	130	684.3	-1.34	0.30	
CO ₂ @Ru-MFI-47	3.3	0.002	44	2.2	125	710.2	-0.34	0.15	
CO2@Rh-MFI-47	3.3	0.002	45	2.28	125	719.7	-0.31	0.22	
CO ₂ @Pd-MFI-47	3.3	0.002	46	2.2	128	804.4	-0.16	0.25	
CO ₂ @Ag-MFI-47	3.3	0.002	47	1.93	134	731	-0.48	0.27	
CO ₂ @Cd-MFI-47	3.3	0.002	48	1.69	148	867.8	-0.78	0.20	
CO ₂ @Ta-MFI-47	3.3	0.002	73	1.5	134	761	-0.51	0.17	
CO ₂ @W-MFI-47	3.3	0.002	74	2.36	130	770	-0.36	0.16	
CO ₂ @Re-MFI-47	3.3	0.002	75	1.9	128	760	-1.14	0.26	
CO ₂ @Os-MFI-47	3.3	0.002	76	2.2	126	840	-0.31	0.15	
CO ₂ @Ir-MFI-47	3.3	0.002	77	2.2	127	880	-0.22	0.21	
CO ₂ @Pt-MFI-47	3.3	0.002	78	2.28	130	870	-0.14	0.29	
CO ₂ @Pb-MFI-47	3.3	0.002	82	1.8	147	715.6	-0.32	0.27	
CO ₂ @Sc-MFI-47	3.6	0.002	21	1.36	144	633.1	-1.04	0.31	
N ₂ @Ti-MFI-47	3.6	0.002	22	1.54	132	658.8	-1.22	0.33	
N ₂ @V-MFI-47	3.6	0.002	23	1.63	122	650.9	-1.32	0.39	
N ₂ @Cr-MFI-47	3.6	0.002	24	1.66	118	652.9	-0.86	0.36	
N ₂ @Mn-MFI-47	3.6	0.002	25	1.55	117	717.3	-1.01	0.34	
N ₂ @Fe-MFI-47	3.6	0.002	26	1.83	117	762.5	-0.70	0.28	

(Supplementary Table 3. Continued)



Supplementary Table 4. Data records in zeolite dataset



Supplementary Figure 9. The adsorption selectivity heat map of (a) CO_2/N_2 , (b) CO_2/H_2O adsorption separation.

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Parameter	value
learning_rate	0.11
max_dapth	16
n_estimators	580
min_child_weight	5
subsample	0.84
colsample_bytree	0.8
reg_alpha	0.8
reg_lambda	0.3

**Supplementary Table 6.** Parameters of the ExtraTree algorithm for predicting the binding energy and adsorption strength index.

Parameter	value
min_samples_leaf	1
max_dapth	12
n_estimators	320
random_state	7
min_samples_split	4



Supplementary Figure 10. The charge density difference of the (a)  $CO_2$ , (b)  $N_2$ , (c)  $H_2O$  on Zr-CHA-47; (d)  $CO_2$ , (e)  $N_2$ , (f)  $H_2O$  on Nb-CHA-47; (g)  $CO_2$ , (h)  $N_2$ , (i)  $H_2O$  on Mo-CHA-47.

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