Inverse Design of ZIFs through Artificial Intelligence Methods.

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1. The ZIFs

	Name	Metal	Linker	Functional Group
1	ZIF-8	Zn	mIm	-CH ₃
2	ZIF-67	Со	mIm	-CH ₃
3	CdIF-1	Cd	mIm	-CH ₃
4	BeIF-1	Be	mIm	-CH ₃
5	Cu-ZIF-8	Cu	mIm	-CH ₃
6	Mg-ZIF-8	Mg	mIm	-CH ₃
7	Mn-ZIF-8	Mn	mIm	-CH ₃
8	ZIF-8-Br	Zn	mIm	-Br
9	Co-ZIF-8-Br	Со	mIm	-Br
10	ZIF-8-Cl	Zn	mIm	-C1
11	ZIF-8-Im_1	Zn	mIm/mIm/Im	-CH ₃ /-CH ₃ /-H
12	ZIF-8-Im_2	Zn	mIm/Im/Im	-CH ₃ /-H/-H

Table S1. The ZIFs of our database.

13	ZIF-8-Im_3	Zn	Im/Im/Im	-H/-H/-H
14	ZIF-7-8	Zn	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-H
15	Co-ZIF-7-8	Со	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-H
16	Be-ZIF-7-8	Be	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-H
17	Cu-ZIF-7-8	Cu	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-H
18	Mg-ZIF-7-8	Mg	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-H
19	Mn-ZIF-7-8	Mn	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-H
20	ZIF-7-8-Cl	Zn	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-Cl
21	ZIF-7-8-Br	Zn	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-Br
22	ZIF-7-8-I	Zn	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-F
23	ZIF-7-8-F	Zn	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-Cl
24	Cd-ZIF-7-8-Cl	Cd	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-Br
25	Cd-ZIF-7-8-Br	Cd	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-Br
26	Co-ZIF-7-8-Cl	Со	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-Cl
27	Co-ZIF-7-8-F	Со	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-F
28	Co-ZIF-7-8-I	Со	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-I
29	Be-ZIF-7-8-F	Be	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-F
30	Be-ZIF-7-8-I	Be	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-I
31	Cu-ZIF-7-8-F	Cu	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-F
32	Cu-ZIF-7-8-Cl	Cu	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-Cl
33	Cu-ZIF-7-8-Br	Cu	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-Br
34	Cu-ZIF-7-8-I	Cu	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-I
35	Mg-ZIF-7-8-Br	Mg	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-Br
36	Mg-ZIF-7-8-I	Mg	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-I
37	Mn-ZIF-7-8-Br	Mn	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-Br
38	Mn-ZIF-7-8-I	Mn	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-I
39	Cd-ZIF-7-8-I	Cd	mIm/mIm/bIm	-CH ₃ /-CH ₃ /-I
40	Tetrz-ZIF-8	Zn	tetrz	-CH ₃

41	Co-Tetrz-ZIF-8	Co	Tetrz	-CH ₃
42	Be-Tetrz-ZIF-8	Be	Tetrz	-CH ₃
43	Cu-Tetrz-ZIF-8	Cu	Tetrz	-CH ₃
44	Tetrz-ZIF-8-NH ₂	Zn	Tetrz	-NH ₂
45	Be-Tetrz-ZIF-8-NH ₂	Be	Tetrz	-NH ₂
46	Co-Tetrz-ZIF-8-NH ₂	Со	Tetrz	-NH ₂
47	dClm-ZIF-8	Zn	dClm	-CH ₃
48	Co-dClm-ZIF-8	Со	dClm	-CH ₃
49	Be-dClm-ZIF-8	Be	dClm	-CH ₃
50	Cd-dClm-ZIF-8	Cd	dClm	-CH ₃
51	Mg-dClm-ZIF-8	Mg	dClm	-CH ₃
52	Cu-dClm-ZIF-8	Cu	dClm	-CH ₃
53	dFm-ZIF-8	Zn	dFm	-CH ₃
54	Co-dFm-ZIF-8	Со	dFm	-CH ₃
55	Cd-dFm-ZIF-8	Cd	dFm	-CH ₃
56	Mg-dFm-ZIF-8	Mg	dFm	-CH ₃
57	Cu-dFm-ZIF-8	Cu	dFm	-CH ₃
58	dIm-ZIF-8	Zn	dIm	-CH ₃
59	Co-dlm-ZIF-8	Со	dIm	-CH ₃
60	Be-dlm-ZIF-8	Be	dIm	-CH ₃
61	Cu-dlm-ZIF-8	Cu	dIm	-CH ₃
62	Mg-dlm-ZIF-8	Mg	dIm	-CH ₃
63	dBrm-ZIF-8	Zn	dBrm	-CH ₃
64	Co-dBrm-ZIF-8	Со	dBrm	-CH ₃
65	Be-dBrm-ZIF-8	Be	dBrm	-CH ₃
66	Cd-dBrm-ZIF-8	Cd	dBrm	-CH ₃
67	Mg-dBrm-ZIF-8	Mg	dBrm	-CH ₃
68	Cu-dBrm-ZIF-8	Cu	dBrm	-CH ₃

69	ZIF-8-CHO	Zn	mIm	-CHO

2. Simulations

2.1. Force fields

The force fields of for the ZIF interactions have been developed with the use of density functional theory computations (DFT) and can be found in our previous work.¹

The force field used consists of the following terms: bond stretching (Eq. 1), bond angle bending (Eq. 2) and torsional angle distortion (Eq. 3) for the bonded intra-molecular interactions, as well as Lennard Jones (LJ) and electrostatic terms, for the non-bonded intra- and inter-molecular interactions (Eq. 4):

$$U^{stretch}(l) = \frac{k_l}{2} (l - l_0)^2$$
(1)

$$U^{bend}(\theta) = \frac{k_{\theta}}{2} (\theta - \theta_0)^2 \tag{2}$$

$$U^{torsion}(\varphi) = k_{\varphi} \left[1 + \cos(m\varphi - \varphi_0) \right]$$
(3)

$$U(r_{ij}) = 4\varepsilon_{ij} \left[\left(\frac{\sigma_{ij}}{r_{ij}} \right)^{12} - \left(\frac{\sigma_{ij}}{r_{ij}} \right)^6 \right] + \frac{1}{4\pi\varepsilon_0} \frac{q_i q_j}{r_{ij}}$$
(4)

where k_l , k_{θ} and k_{φ} are constants that characterize the bond length, bond angle and torsional angle stiffness, respectively; l, θ and φ correspond to the bond length, bond angle and torsional angle, respectively, and subscript 0 refers to the equilibrium value; ε_{ij} and σ_{ij} are the Lennard-Jones energy and size parameters, r_{ij} is the distance between atoms *i* and *j*, q_i is the charge at atom *i* and ε_0 is the permittivity of vacuum.

The force-field development has presented in our previous work.¹

2.2. Simulations in the ZIFs

Each modification was reconstructed on the molecular level, in the form of a super-cell, which corresponds to a box of $(2\times2\times2)$ unit cells. Then, each ZIF variant underwent equilibration with Molecular Dynamics (MD) simulations, at the NPT ensemble, at 308 K and 1 bar for 1 ns, in order to allow for correct framework volume adjustment. The importance of applying an MD simulation at the NPT ensemble prior to the main simulations should be underlined because each new replacement unit affects the framework's volume, which in turn can affect the resulting aperture size. Thus, considering a common volume value across the different ZIF simulation boxes would be incorrect. Then, the structures underwent a similar thermal equilibration, at the NVT ensemble (again at 308 K) for 1 ns. The thermostat in all simulations was Nose-Hoover and the time step was set to 1fs.

2.2.1. dcTST

Conventional MD methods cannot predict diffusivities lower than 10⁻¹² m²/sec,² and the nonconventional approach of dcTST had to be used in our case, where in a lot of systems the observed diffusivities were considerably below than this limit. In dcTST diffusivity is approached as a success of "hops" of a molecule under investigation from cage to cage, by crossing the framework's apertures. Numerous attempts lead to a successful aperture crossing, and the success rate can be translated to the diffusivity, through the following expression:

$$D_0 = \frac{1}{2n} k_{EXIT} l^2 \tag{5}$$

where *n* depends on the dimensionality of diffusion, which in ZIF-8 equals to 3, k_{EXIT} is the total exiting rate of a molecule from a cage to any of the adjacent ones, and *l* is the distance between energy minima, which are the centers of cages (for example, this distance in ZIF-8 is 14 Å).

The total exiting rate is estimated by:

$$k_{EXIT} = n_{apertures} k_{crossing} = 8k_{crossing}$$
(6)

Where $n_{apertures}$ is the number of available exiting apertures in a cage (8 in SOD topology) and $k_{crossing}$ is the rate of successful crossings of a molecule through a given aperture, through

$$k_{crossing} = \frac{1}{\sqrt{2\pi m}} P(\lambda^*) \tag{7}$$

where m is the mass of the molecule under investigation, λ is the reaction coordinate and can be regarded as a function of the Cartesian coordinates (see Figure S3(b)). $P(\lambda^*)$ is the probability of finding the molecule in the dividing surface, which is an orthogonal plane at $\lambda = \lambda^*$ vertical to the reaction axis, close to the energy barrier (or at the point free energy is maximized). In our case, the energy barrier corresponds to the aperture's center. The probability $P(\lambda^*)$ depends on the free energy barrier and is obtained from:

$$P(\lambda^*) = \sqrt{\frac{k_B T}{2\pi m}} \frac{e^{-\beta F(\lambda^*)}}{\int_{-\infty}^{\infty}} e^{-\beta F(\lambda)}$$
(8)

where *F* stands for the free energy and it is provided by the umbrella sampling analysis discussed above.

The crossing rate, $k_{crossing}$, of equation 7 does not account for some seemingly successful jumps failing to thermalize in the new cage that end up back again to the cage they came from.³ To account for this, a dynamic correction factor must be calculated, called correction factor (κ). Thus, the actual, crossing rate is:

$$k_{crossing}^{\ \ dc} = \kappa k_{crossing} \tag{9}$$

The probability, $P(\lambda^*)$, in equation 8, is estimated with the help of umbrella sampling.⁴ Details on the umbrella sampling and the estimation of the correction factor can be found in our previous work.⁵

2.2.2. Solubilities with test particle insertion

In test particle insertion a single molecule is introduced in a ZIF, at different directions and positions, in numerous ZIF configurations that have been produced beforehand through MD simulations, in the isochoric-isothermal (N, V, T) ensemble. Then, the excess chemical potential of component *i* is calculated through:⁶

$$\mu_{i}^{ex} = -\frac{1}{\beta} \ln \left[\left\langle \exp\left(-\beta U_{test}^{inter}\right) \right\rangle_{Widom,N,V,T} \right]$$
(10)

where $\beta = 1/k_B T$ and U_{test}^{inter} is the intermolecular energy of the inserted gas molecule, due to its interaction with the surrounding ZIF atoms. The averaging indicated by the brackets is carried out over all the degrees of freedom of the inserted gas molecule and all ZIF conformations. For each ZIF under study, 1000 different structure configurations were produced with MD simulations and in each configuration a single gas molecule is inserted 50,000 times. The solubility, S_i , of species *i*, is given by:⁶

$$S_{i} = \frac{22400 cm^{3}(STP) \ 1}{mol \ RT} exp\left(-\frac{\mu_{i}^{ex}}{RT}\right)$$
(11)

Finally, the permeability of species *i* can be calculated through:

$$P_i = D_i \times S_i \tag{12}$$

3. AI tools development: Machine Learning and Genetic Algorithms

3.1. Table with all available building units considered in our algorithm

The descriptors of the ML models were based on readily-available information about the linkers, functional groups and the metal center of each ZIFs, as well as information about the guest molecules, that does not require additional calculations from a potential user of our tool, such as potential energy surfaces.^{7,8} Sixteen descriptors (ML features) were used for ZIFs (Table S2): linker length (×3), linker mass (×3), and ε and σ of the outermost atoms of the linker forming the apertures bridging the cages; the length and mass of the functional groups (×3); the radius and the atomic number of the metal. For the gas molecules (Table S3), the descriptors were: van der Waals and kinetic diameter, mass and acentric factor.

Building unit	Descriptors
Metal	Atomic Num.
Wetar	Radius
Linker 1	Length
	Mass

Table S2. ZIF descriptors in our ML models.

	Sigma
	Epsilon
	Length
Linker 2	Mass
	Sigma
	Epsilon
	Length
Linker 3	Mass
	Sigma
	Epsilon
Fune group 1	Length
Func. group 1	Mass
Func aroun 2	Length
r une. group 2	Mass
Euro mouro 2	Length
Tune. group 5	Mass

Table S3. Gas descriptors in our ML models.

Descriptors van der Waals diameter Kinetic diameter Mass Ascentric factor

Table S4. Linkers

	Length (Å)	Mass (u)	σ (Å)	e (kJ/mol)
Imidazolate	4.438	81	0.25	0.0627
bIm	5.996	117	0.25	0.0627
Tetrazolate	3.66	83	0.325	0.7112
dClm	5.7	134.9	0.34	1.2552
dFm	4.86	101.98	0.285	0.255
dBrm	6.010	223.8	0.4	1.8731

dIm	6.410	317.8	0.367	2.4267

Table S5. Functional groups

	Length (Å)	Mass(u)
-CH ₃	3.78	15
-Br	3.85	79.9
-Cl	3.54	35.45
-H	2.278	1
-NH ₂	3.927	16
-СНО	4.093	31

Table S6. Metals

	Ionic Radius (pm)	Metal Number	Metal Mass (u)
Cd ²⁺	92	48	112.41
Zn ²⁺	74	30	65.38
Cu ²⁺	71	29	63.456
Co ²⁺	72	27	58.93
Be ²⁺	41	4	9.012
Mg	71	12	24.305
Mn	80	25	54.938

3.2. Details on the gas molecules considered.

Table S7. The gas molecules considered, along with their properties.

Mass vdW diameter⁹ Kinetic diameter⁹ Acentric factor¹⁰

	(g/mol)	(Å)	(Å)	
He	4.00	2.66	2.60	-0.390
H_2	2.01	2.76	2.89	-0.217
O_2	32.00	2.94	3.46	0.022
CO ₂	44.01	3.24	3.30	0.225
N_2	28.00	3.13	3.64	0.037
CH ₄	16.04	3.25	3.80	0.011
C_2H_4	28.05	3.59	3.90	0.087
C_2H_6	30.07	3.72	4.00	0.099
Kr	83.8	4.00	3.69	0.005
Xe	131.29	4.1	4.1	0.008
C ₃ H ₆	42.08	4.03	4.50	0.142
C_3H_8	44.10	4.16	4.30	0.152
Rn	222	4.2	4.2	0.008
n-C ₄ H ₁₀	58.12	4.52	4.50	0.200
i-C ₄ H ₁₀	58.12	4.80	4.42	0.183

3.3. Details on the ML models

3.3.1. The Dataset

The dataset used in the training of the ML models is an extended version of a dataset we published recently.¹ It consists of ZIFs that we manually designed and equilibrated with molecular simulations, with force fields developed in house. More specifically:

- 1. The training dataset in the present publication misses the three designed ZIFs (for the three separations). These ZIFs with the guidance of our GA tool in the original dataset.
- 2. The training dataset in the present publication is extended in the sense that we expanded the range of the gas molecules. We carried out simulations for the diffusivity calculation for Kr, Xe and Rn (additionally to the gas molecules included in the original dataset).

3.3.2. The ML models

Scikit learn was used for the ML models. The parameters for the various regressors are supplied in Table S8.

Table S8. Information on the ML regressors of our work.

Regressor type	Parameters
LR	None
DT	<i>Max depth</i> : 6
RF	Max depth: 6, $n_{estimators} = 600$
NN	<i>Layers</i> : (50, 30, 30, 20, 10); <i>solver</i> : lbfgs; <i>max_iter</i> = 5000
GP	Kernel = ConstantKernel(1.0, (10 ⁻² , 10 ³))×DotProduct(5, (10 ⁻⁷ , 10 ³)) ×RBF(5, (10 ⁻⁷ , 10 ³))
	M_restarts_optimizier=20, alpha=0.01, normalize_y=True
GBR	n_estimators = 500; learning_rate = 0.2; <u>max_depth</u> =6; loss='squared_error'
XGBR	n_estimators=500; max_depth=5; eta=0.07; subsample=0.75, colsample_bytree=0. 7, reg_lambda=0.4, reg_alpha=0.13

The performance of the Ml models, as shown in the plot of Figure 1, in the main text, was accomplished by using the *K*-fold cross-validation protocol:¹¹ the set is split randomly in *K* folds, that do not overlap. One of the folds is the test set, and the rest are used to train the models. This procedure is repeated, until all folds have served as the test set. The metrics used were R^2 , med absolute (ABS) error, mean squared error (MSE), the maximum error and mean ABS perc error. The performance is shown in the following Table:

	R ²	Med ABS error	Max error	Mean ABS error	MSE
LR	0.761	1.726	19.181	0.167	7.861
DT	0.890	0.851	10.049	0.089	3.635
RF	0.925	0.764	6.857	0.075	2.485
NN	0.919	0.610	9.092	0.072	2.680

Table S9. Performance metrics of the various ML models developed in this work.

GBR	0.984	0.337	5.442	0.037	0.550
XGBR	0.985	0.290	6.5	0.030	0.528

3.4. Details on our genetic algorithm

3.4.1. Genetic algorithm development

The development of the genetic algorithm was based on the PyGAD library.¹²

Parent selection

For the parent selection we chose the K tournament method, while the crossover type was set to uniform. Six (6) parents mated at each generation, which were selected through the K tournament method, with K tournament = 12.

K_tournament = 12

The 4 best candidates at the end of each generation run were kept in the population (keep_eliticism). The crossover probability was 0.7.

Mutation

Most importantly, we used the adaptive mutation,¹³ which avoids the pits of the total randomness of mutation, by promoting the evolution of high quality genes and discouraging the transfer of "bad" genes to the next generations. The mutation probabilities were 0.6 for solutions whose performance was below the population average performance, and 0.06 for solutions whose performance was above the population average performance.

Number of generations

In the case of the O_2/N_2 , 3000 generation steps were used. For the design of optimum ZIF for CO_2/CH_4 and C_3H_6/C_3H_8 , where our algorithm's design was limited to symmetrical ZIFs, the number of generation steps was reduced to 100.

3.4.2. The three ZIFs

In the case of O_2/N_2 , the GA algorithm could freely construct ZIFs of independent linker and functional group composition. In the case of ZIFs designed for CO_2/CH_4 and C_3H_6/C_3H_8 separations, we limited the ZIF generation routine in our GA algorithm, to construct symmetrical ZIFs (linker1=linker2=linker3; functional group 1=functional group 2=functional group 3),

because the search space gets rapidly crowded by proposed optimized structures for the given performance boundaries.



Figure S1. New solutions rate as a function of the generation number for the case of (a) O_2/N_2 , (b) CO_2/CH_4 , and (c) C_3H_6/C_3H_8 .

ZIF	gas	D	S	Permeability
		(m^2/s)	(cm ³ (STP)/cm ³ cmHg)	(Barrer)
Cd-I-ZIF-7-8	O ₂	1.2×10 ⁻¹²	5.3×10-2	6.57
	N_2	5.8×10 ⁻¹⁴	5.1×10 ⁻²	0.12
dFmBe	CO ₂	6.6×10 ⁻¹⁰	0.5	1.1×10^{4}
	CH_4	1.0×10 ⁻¹⁴	0.24	0.24
ZIF-67	C_3H_6	2.0×10 ⁻¹³	5.2	154.71
0,	C_3H_8	1.5×10 ⁻¹⁵	9.2	1.38

Table S10. Simulation results

membrane	$P_{\rm CO2}$ (Barrer)	P_{CO2}/P_{CH4}
ZIF-8 ¹⁴	7.16×10 ³	5.1
ZIF-8 (31%)/P84 ¹⁵	20	95
Matrimid [®] -PEG ¹⁶ /ZIF-8 (4%)	27.5	24
PDMS/ACN-O (10%) ¹⁷	4.3×10^{3}	4.5
PVC-g-POEM ¹⁸	43.5	18.1
PVC-g-POEM/H-ZIF 5% ¹⁸	51.7	17.2
PVC-g-POEM/H-ZIF 10% ¹⁸	210.6	14.3
PVC-g-POEM/H-ZIF 20% ¹⁸	224.7	11.9
P84/ND ¹⁹	1.61	75
Matrimid/ZIF-1 (10%) ²⁰	6.75	23
6FDA-TTM/Si-H (5%) ²¹	29.7	76
MWCNT@GONRs (2%) ²²	25.2	11
LOYMET ²³	29	31.74
KIOJUQ ²³	478	22.34
15wt% ZIF-90B/6FDA-DAM ²⁴	720	37
$PIM-EA(H_2)-TB^{25}$	1.4×10^{3}	17.7
MOF@COF/PSf-5 MMM ²⁶	7	50
NH2-MIL-53(Al) ²⁷	53	28.7
PVC-g-POEM/H_ZIF-8 ²⁸	62.3	11.2
SSZ-13 ²⁹	6×10^{3}	300
Matrimids-CNTs/GO-5/5 ³⁰	38	85
VAD-OS-SAPO-34 ³¹	4.3×10^{3}	158
Matrimid/UiO-67-33 ³²	27	75
6FDA-DAM/UiO-66-COCH ₃ ³³	1.2×10^{3}	33
Pebax [®] -P84 ^{®34}	19	114
Pebax [®] -P84 [®] -ZIF-8 ³⁴	19	65
Pebax [®] -P84 [®] -MIL-101 ³⁴	19	60
Pebax [®] -P84 [®] -UiO-66 ³⁴	29	56
Pebax®-P84®-ZIF-7/8 ³⁴	18	50
CHA S-L-18 ³⁵	2×10^{3}	30
PIM-BTrip ³⁶	1×10^{4}	25.6
PIM-DM-BTrip ³⁶	1.3×10^{4}	20
PIM-DTFM-BTrip ³⁶	1.7×10^{4}	18.7
PIM-DM-BTrip ³⁶	16×10^{4}	16.5
PIM-DM-BTrip ³⁶	23×10^{4}	13.2
PIM-TMN-Trip ³⁶	23×10^{4}	15.7
PIM-TFM-BTrip ³⁶	33×10 ⁴	14.7
Pebax/ZIF-8 ³⁷	118	21
Pebax/ZIF-67 ³⁷	162	25
ZIF-7 ³⁸	5.6	13
IRMOF-1 ³⁹	1.1×10^{4}	328

Table S11. Literature data in Figures 1 and 3(a) (CO₂/CH₄)

DD3R ⁴⁰ (zeolite)	3.8×10^{3}	590
SAPO-34 ⁴¹ (zeolite)	2.99×10 ³	115
ALPO-18 ⁴² (zeolite)	1.97×10^{3}	59.2
T ⁴³ (zeolite)	2.75×10^{3}	400
ZIF-62 glass ⁴⁴	2.64×10^{3}	36
CAU-1-NH ₂ /organosilica ⁴⁵	1.07×10^{5}	18
SIM-146	2.6×10^{3}	1.1
Bio-MOF-1 ⁴⁷	5.4×10^{4}	2.6
Sod-ZMOF-1 ⁴⁸	94	3.6
Bio-MOF-1 ⁴⁹	16.57	42.6
MnP3nF7 ⁵⁰	16.48	82.4
Matrimid/NiDOBDC/GO ⁵¹	10	58.3
PSM-MOF ⁵²	1.89×10^{3}	17.7
PERA/CuBTC ⁵³	96.2	23.1
Polymer/NH ₂ -ZIF-8 ⁵⁴	218	13.84
Pehax®1657 UiO-66-NH ⁵⁵	97 5	22.1
Pebax®1657 UiO-66 ⁵⁵	118 3	30.5
$Pebax @ 1657 7 IF - 8^{56}$	758	16.1
Pebax \mathbb{R} 1657 CuBTC ⁵⁷	56.2	23.4
Pehax®7533 7IF-1158	402.89	12 49
Polysulfone 7IF-8 ⁵⁹	12.05	19.8
Polysulfone Mn(HCOO) ₂ ⁶⁰	6.83	9.16
Polysulfone $Cu_2(BTC)_2^{60}$	304.4	3.6
Polysulfone MIL -10161	32.0	23 50
$\frac{101}{2} = \frac{101}{101}$	120	23.30
Pehav $(R_{1657} \text{ NH}_{-53}(\Lambda))^{62}$	129	20.5
1000000000000000000000000000000000000	95 7	12 9
$\frac{1}{2} \frac{1}{2} \frac{1}$	101	12.9
$\frac{1}{2} \frac{1}{2} \frac{1}$	130	12.0
Pehav MH 1657 a -Ni(im)-64	100.6	33 /
P_{ebax} (1657/SAPO-34 (50%) ⁶⁵	338	ларана 16
Pehav/UiO_66_NH_66	845	10 20
1000000000000000000000000000000000000	562 5	20 16
Pebay/UiO_66_(COOH)_66	600	17.5
PSf/NHMII -53(A1)67	6	17.5
$PSf/NH_{-}MII_{-53}(A1)^{68}$	4.5	
$PVDF/NHMII53(A1)^{69}$	1.3	20
$PI/NH_{2}-NII_{2}-33(AI)$	2 5	20.03
$PMP/NHMII53(A1)^{70}$	330 5	דד 22 8
$PI/NH_{2}-NIL-53(A1)^{71}$	0.2	22.0
$\frac{11}{112} \frac{112}{112} 112$	3	2.1
Matrimid/NH MIL 52(A1) ⁷²	5	30.1
Matrimid/ NH ₂ -WIL-55(AI) ⁷²	1.5	42
$\frac{1}{1} \frac{1}{1} \frac{1}$		19.2
$\frac{1}{1} \frac{1}{1} \frac{1}$	2.0 2.9	δ <i>2</i> 04
$\frac{1}{1000} \frac{1}{1000} \frac{1}{1000} \frac{1}{1000} \frac{1}{1000} \frac{1}{1000} \frac{1}{10000} \frac{1}{10000000000000000000000000000000000$	5.0 1.7	94 120
1VIatrimid/ 1 MH ₂ - 1 VIIL- 33 (AI) ⁷³	1./	139

Matrimid/ZIF-7 ⁷⁴	4.5	155
Matrimid/ Ni ₂ (dobdc) ⁷⁵	14.7	32.5
Matrimid/ZIF-8 ⁷⁶	12.96	41.5
Pebax 1675/CuNi ⁷⁷	70.9	18.65
PEEK-WC/CuNi ⁷⁷	47.84	47.84
6FDA-mPD/MOF-199 ⁷⁸	50	89
6FDA-BI ⁷⁹	9	75.3
P/Zn^{2+79}	9.1	79.8
6FDA-BI/20%ZIF-8 ⁷⁹	20.3	57.9
6FDA-BI/20%ZIF-8 0.004 Zn ²⁺⁷⁹	22.8	60.3
6FDA-BI/20%ZIF-8 0.007 Zn ²⁺⁷⁹	27.4	56.1
6FDA-BI/20%ZIF-8 0.01 Zn ²⁺⁷⁹	15.9	70.2
D84 [N; (HCOO) 180	1	62
r 84 - [INI3(IICOO)6]	1.3	67
PN10 ⁸¹	96.2	22.9
PN20 ⁸¹	145	25.4
PN30 ⁸¹	264	27.9
PN40 ⁸¹	395	36.3
PPO/ZIF-8 (3 wt%) ⁸²	76.1	18.1
PPO/ZIF-8 (6 wt%) ⁸²	75.3	17.7
PPO/ZIF-8 (10 wt%) ⁸²	99.5	17.3
PPO/ZIF-8 (15 wt%) ⁸²	114.1	17.3
PPO/ZIF-8 (25 wt%) ⁸²	189	15.9
PPO/ZIF-8 (35 wt%) ⁸²	314.2	16.6
PPO/ZIF-8 (45 wt%) ⁸²	448.7	11.8
ZIF-S ⁸³	2802	21.01
ZIF-S ⁸³	3390.7	16.92
ZIF-S ⁸³	3762.34	17.35
ZIF-S ⁸³	3809	14.23
ZIF-S ⁸³	4219.9	12.62
ZIF-S ⁸³	4517.1	12.5
Pebax ZIF-8 3% ⁸⁴	169.719	19.105
Pebax Zif-8 10% ⁸⁴	174.37	19.187
Pebax ZIF-8 20% ⁸⁴	198.55	15.056
PIM-1/ZIF-7 ⁸⁵	3640	16.6
PIM-1/NH ₂ -ZIF-7 ⁸⁵	2948	20.7
PMP/Pebax/CNF-UiO ⁸⁶	186.9	16.5
PMP/Pebax/CNF-UiO ⁸⁶	220.4	21.1
CuBTC-PVDF 0%87	0.9115	21.27
CuBTC-PVDF 5% ⁸⁷	1.067	24.81
CuBTC-PVDF 10%87	2.002	41.7
CuBTC-PVDF 15%87	3.206	40.07
CuBDC-PVDF 5% ⁸⁷	1.126	26.18
CuBDC-PVDF10%87	1.602	35.6
CuBDC-PVDF15% ⁸⁷	1.987	45.15
MIL-53(Al) 5% ⁸⁷	1.21	21.22
- () -		

MIL-53(Al) 10%87	1.553	20.98
NH2-MIL-53(Al) 5%87	1.107	23.06
NH2-MIL-53(Al) 10%87	1.406	26.03
MIL-101(Cr)-SPEEK ⁸⁷	31	31
MIL-101(Cr)-SPEEK ⁸⁷	29	29
MIL-101(Cr)-SPEEK ⁸⁷	1623	33
MIL-101(Cr)-SPEEK ⁸⁷		
Commercialy attractive region by:		
Hilock et al. ⁸⁸	33.7	35.1

Table S12. Literature data in Figure 3(b) (O_2/N_2)

membrane	P_{O2} (Barrer)	$P_{\rm O2}/P_{\rm N2}$
ZIF-8 ⁸⁹	7.32×10 ³	2.71
ZIF-8-poly(1,4-phelyene ether-ether- sulfone) ⁹⁰	10.65	5.33
Polyurethane – ZSM-5 ⁹¹	21.5	2.7
PIM-SBF ⁹²	1.87×10^{3}	3.4
PIM-1 ⁹³	600	4.8
$PSF - P (0.5 \ \mu m)^{94}$	30	1.8
$PSF - P (10 \ \mu m)^{94}$	70	7
PSF-3PDMS ⁹⁵	82.5	4.6
PIM-1A/NanoMIL-10196	3×10 ³	3
Matrimid®5218/PIM-EA(H ₂)-TB ²⁵	350	5.6
Matrimid®-ZIF-897	5.6	6.4
poly(Pn4) ⁹⁸	494	2.2
PIM-TMN-Trip ³⁶	152	6.4
PIM-HMI-Trip ³⁶	706	4.9
PIM-DM-BTrip ³⁶	1.3×10^{3}	4.5
PIM-DTFM-BTrip ³⁶	1.9×10^{3}	3.5
PIM-TFM-BTrip ³⁶	3×10 ³	3.8
	63	6.5
2abIm-VPLT-LIPS-ZIF-899	111	5.7
	218	5.9
Sod-ZMOF-1 ⁴⁸	18.5	1.7
PSM-MOF ⁵²	310	2.84
Pebax 1675/CuNi ⁷⁷	3.8	2.71
PEEK-WC/CuNi ⁷⁷	6.51	4.93
Pebax ZIF-8 3% ⁸⁴	7.261	1.791
Pebax Zif-8 10% ⁸⁴	8.929	2.109
Pebax ZIF-8 20% ⁸⁴	11.314	2.349
MOF-801/PIM-1 ¹⁰⁰	1752	4.8
Commencially attractive maximum have		

Commercially attractive region by:

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