

SUPPLEMENTARY

NEURAL-NETWORK DEEP POTENTIAL MODELS FOR LITHIUM ADSORPTION ON
C–N–O-MODIFIED CARBON MATERIALS

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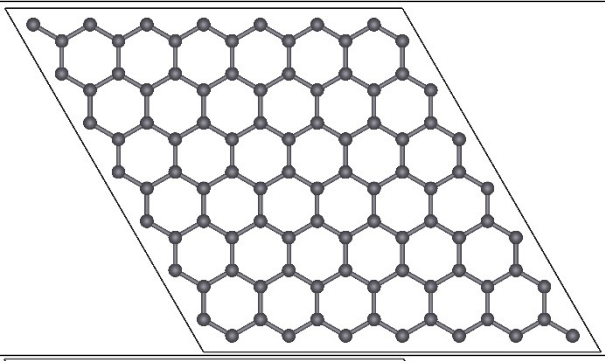
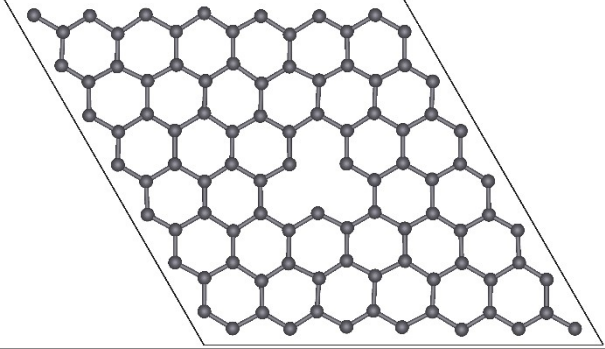
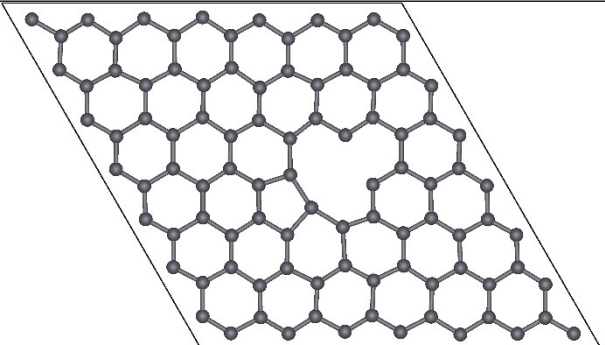
South Ural State University, 454080 Chelyabinsk

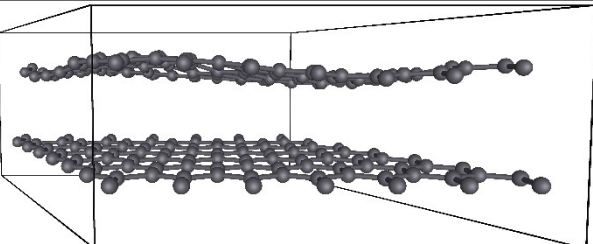
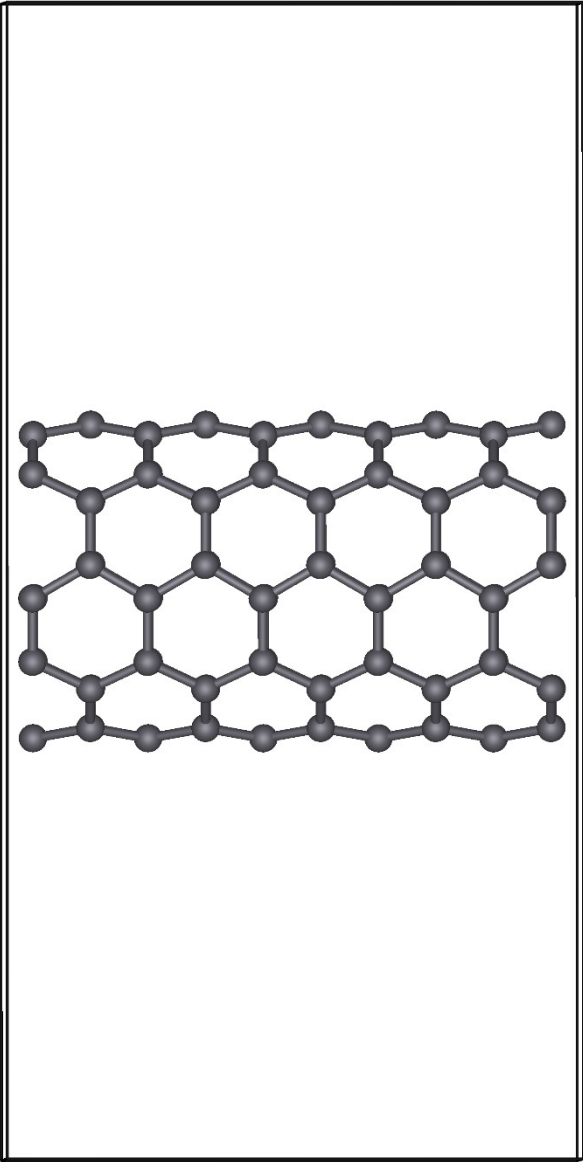
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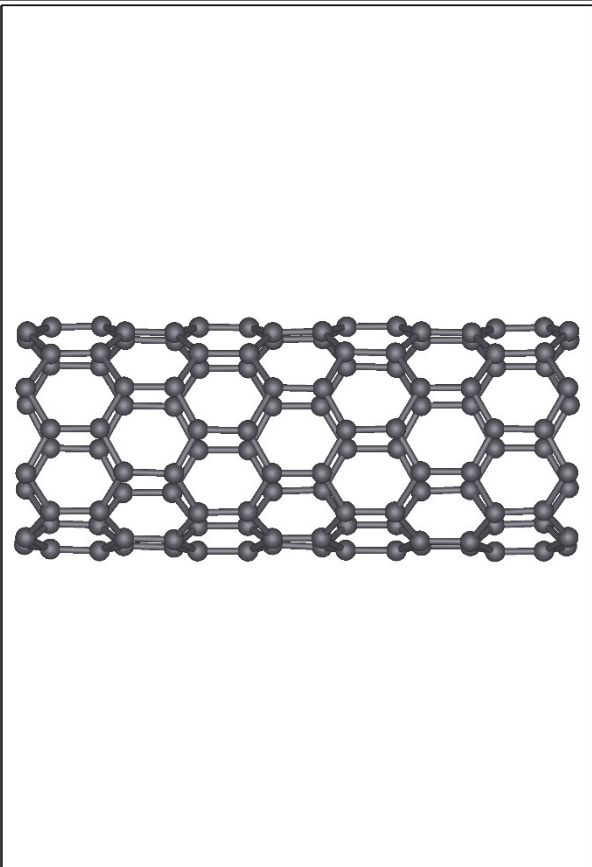
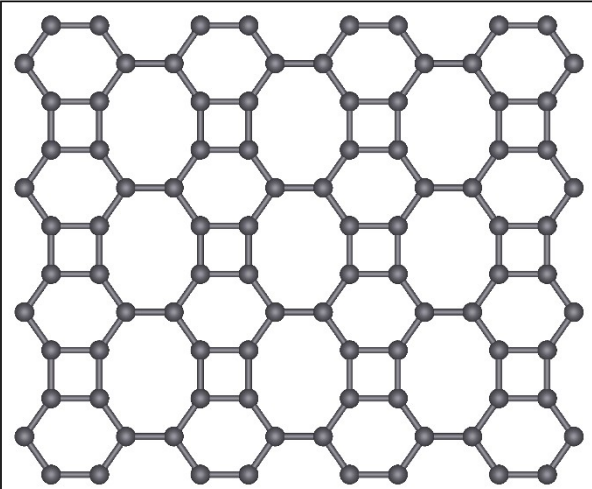
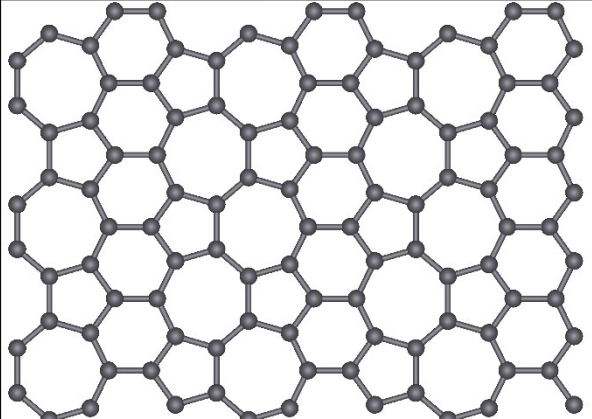
Dataset composition

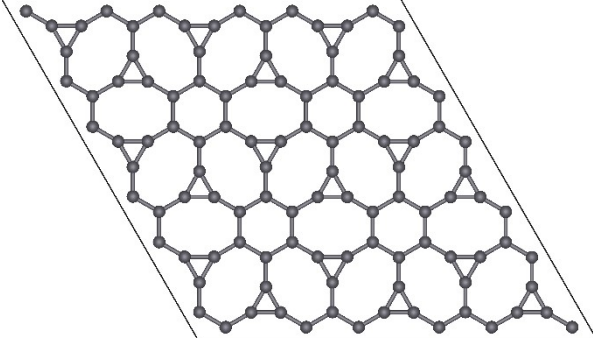
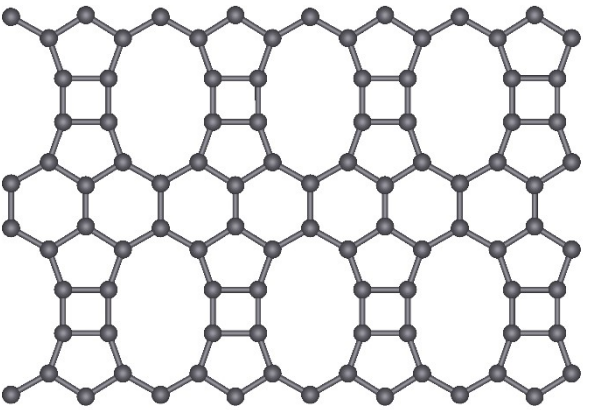
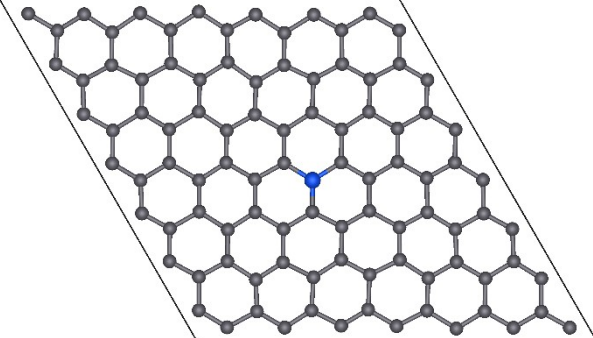
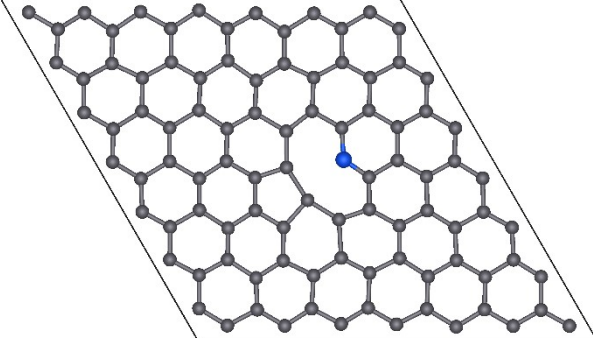
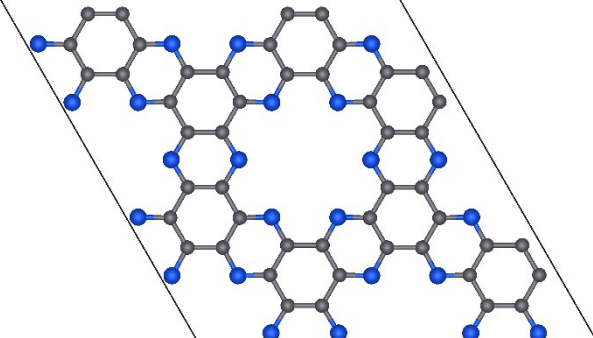
The table S1 summarises the different system types included in the dataset, along with a representative atomic structure for each type and the number of corresponding configurations. For equilibrium structures, the simulation cell parameters are also provided.

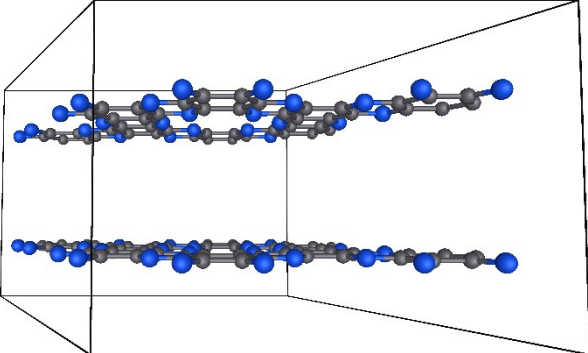
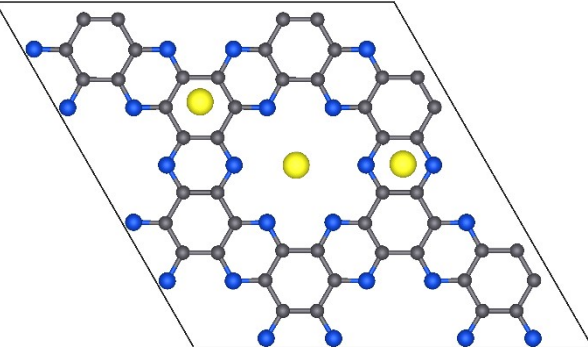
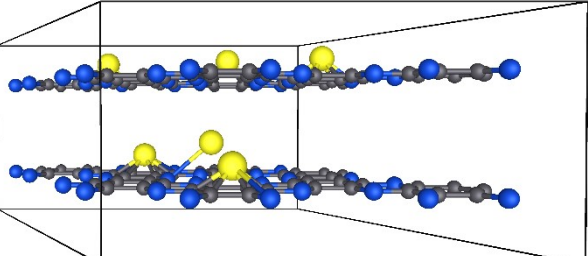
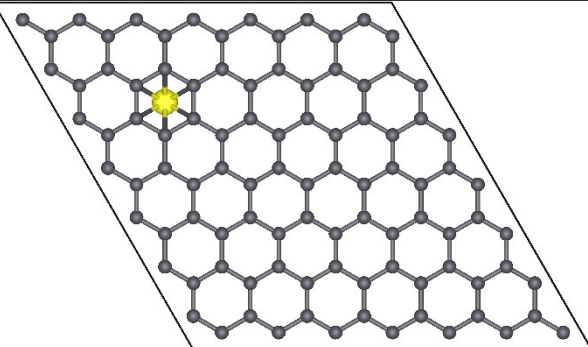
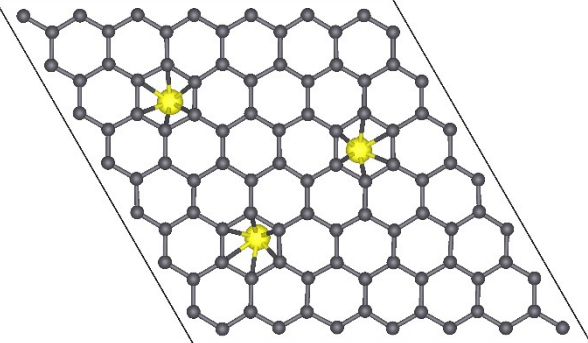
Table S1. Composition of the reference dataset

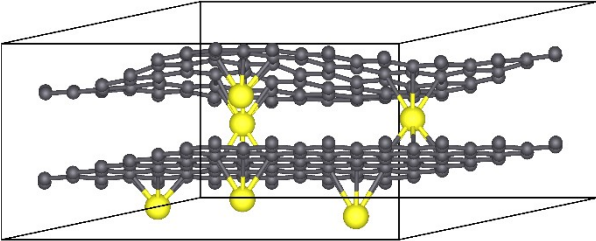
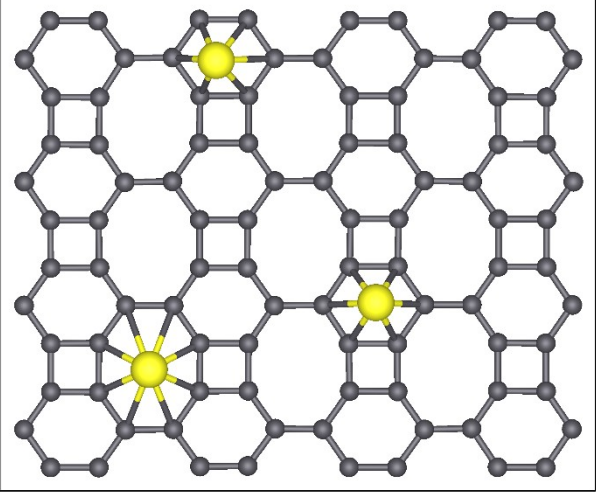
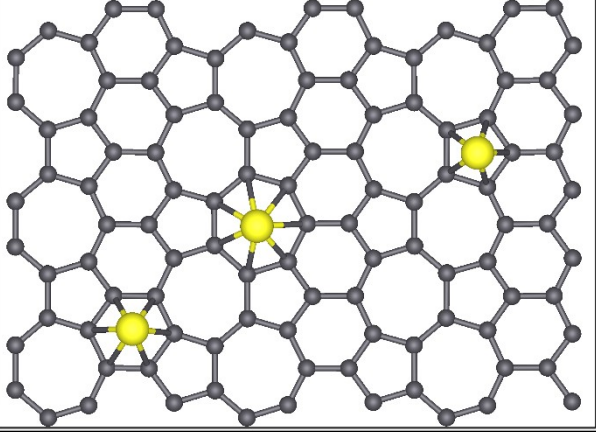
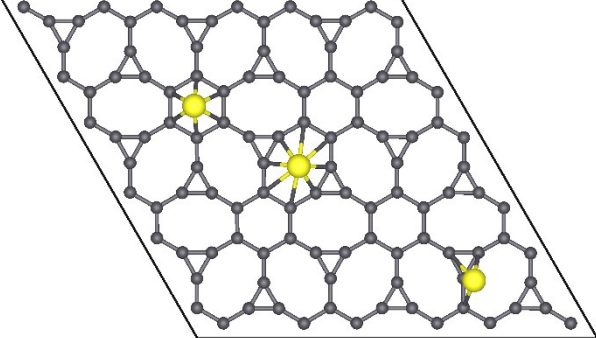
System type, cell parameters (for stable models)	Structure	Number of structures in dataset
graphene, a=17.27, b=17.27		282
graphene_vac – graphene with vacancy defect		89
graphene_dvac – graphene with double vacancy defect		61

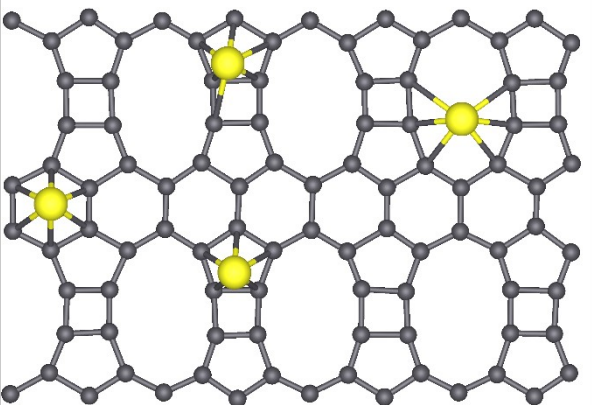
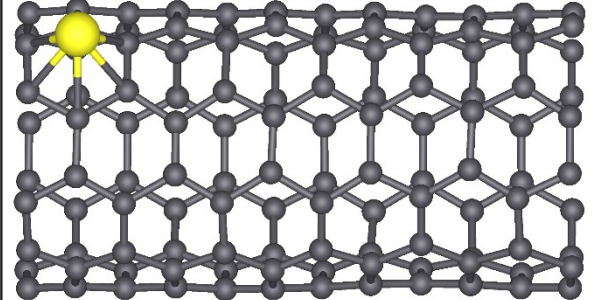
<p>graphite</p>		<p>480</p>
<p>SWNT(5,5) – single-walled carbon nanotube (5,5), $c=12.33$</p>		<p>1339</p>

<p>SWNT(8,0) – single-walled carbon nanotube (8,0), c=17.04</p>		<p>1609</p>
<p>biphenylene, 18.08 15.05 25.11</p>		<p>139</p>
<p>psi-graphene, a=20.12, b=14.51</p>		<p>171</p>

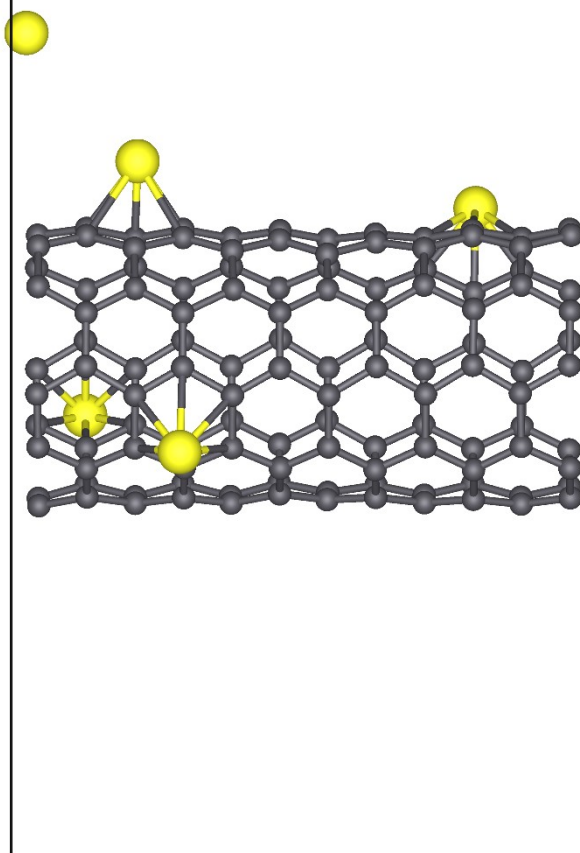
<p>irida-graphene, a=18.95, b=18.95</p>		<p>161</p>
<p>tpdh-graphene, a=19.77, b=13.94</p>		<p>110</p>
<p>N-graphene – graphene with a substitutional nitrogen defect</p>		<p>387</p>
<p>N-pyridinic – graphene containing pyridinic nitrogen</p>		<p>433</p>
<p>C_2N, a=16.65, b=16.65</p>		<p>2159</p>

<p>C_2N_3D – stacked C_2N structure with three-dimensional periodicity</p>		<p>337</p>
<p>C_2N+3Li – C_2N with three adsorbed Li atoms</p>		<p>1566</p>
<p>C_2N_3D+3Li – stacked C_2N structure with three-dimensional periodicity and three adsorbed Li atoms per C_2N layer</p>		<p>48</p>
<p>graphene+1Li – graphene with one adsorbed Li atom, $a=17.28$, $b=17.28$</p>		<p>271</p>
<p>graphene+3Li – graphene with three adsorbed Li atoms</p>		<p>292</p>

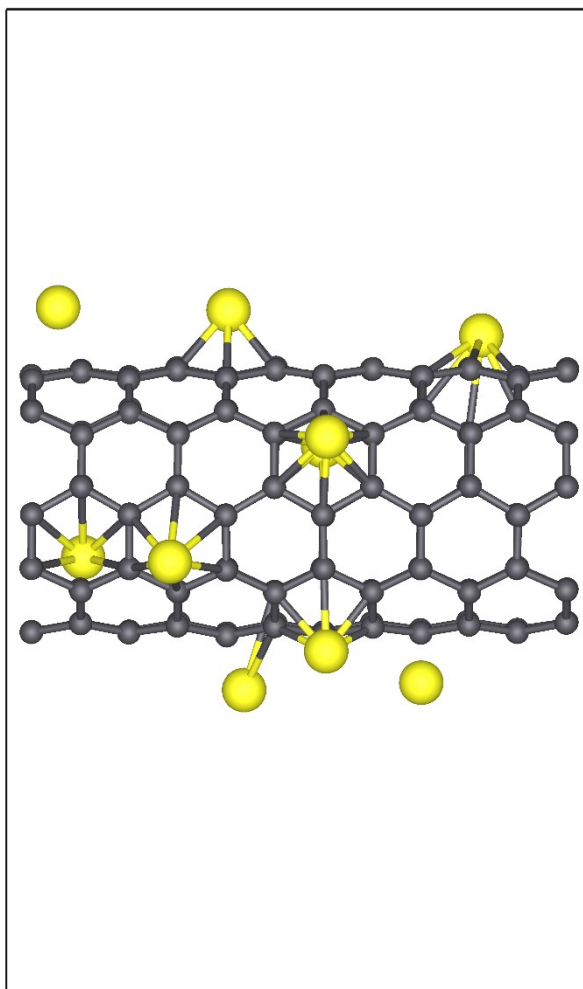
<p>graphite+3Li – graphite with three adsorbed Li atoms per graphene layer</p>	 <p>A 3D perspective view of a graphite crystal structure, showing multiple stacked layers of carbon atoms. Three yellow spheres representing lithium atoms are shown adsorbed on the surface of one of the layers.</p>	<p>675</p>
<p>biphenylene+3Li – biphenylene with three adsorbed Li atoms</p>	 <p>A 2D top-down view of a biphenylene molecule, which consists of four fused benzene rings. Three yellow spheres representing lithium atoms are shown adsorbed on the surface of the molecule.</p>	<p>389</p>
<p>psi-graphene+3Li – psi- graphene with three adsorbed Li atoms</p>	 <p>A 2D top-down view of a psi-graphene lattice, which is a honeycomb-like structure of carbon atoms with a different arrangement than standard graphene. Three yellow spheres representing lithium atoms are shown adsorbed on the surface.</p>	<p>423</p>
<p>irida-graphene+3Li – irida-graphene with three adsorbed Li atoms</p>	 <p>A 2D top-down view of an irida-graphene lattice, which is a honeycomb-like structure of carbon atoms with a different arrangement than standard graphene. Three yellow spheres representing lithium atoms are shown adsorbed on the surface.</p>	<p>570</p>

<p>tpdh-graphene+4Li – tpdh-graphene with four adsorbed Li atoms</p>		<p>435</p>
<p>SWNT(5, 5)+1Li – SWNT(5, 5) with one adsorbed Li atom</p>		<p>658</p>

SWNT(5, 5)+5Li –
SWNT(5, 5) with five
adsorbed Li atoms

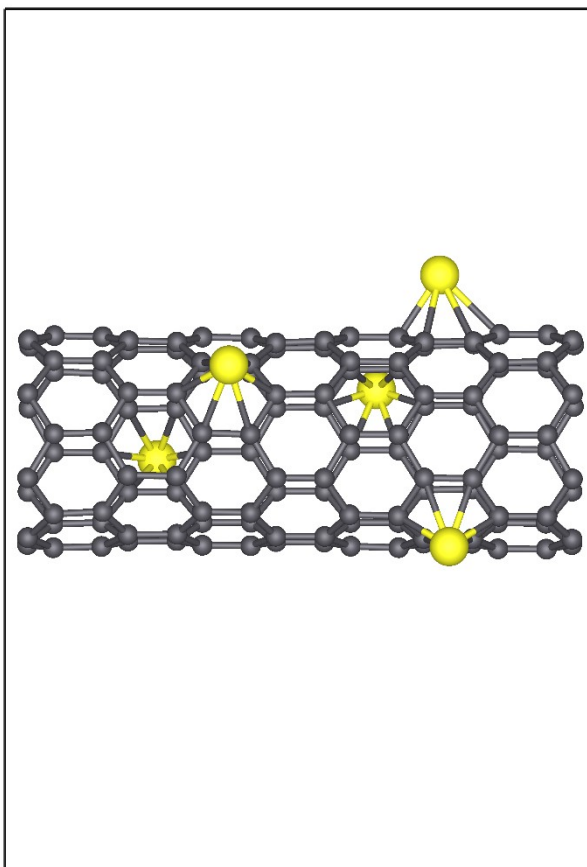


SWNT(5, 5)+10Li –
SWNT(5, 5) with ten
adsorbed Li atoms



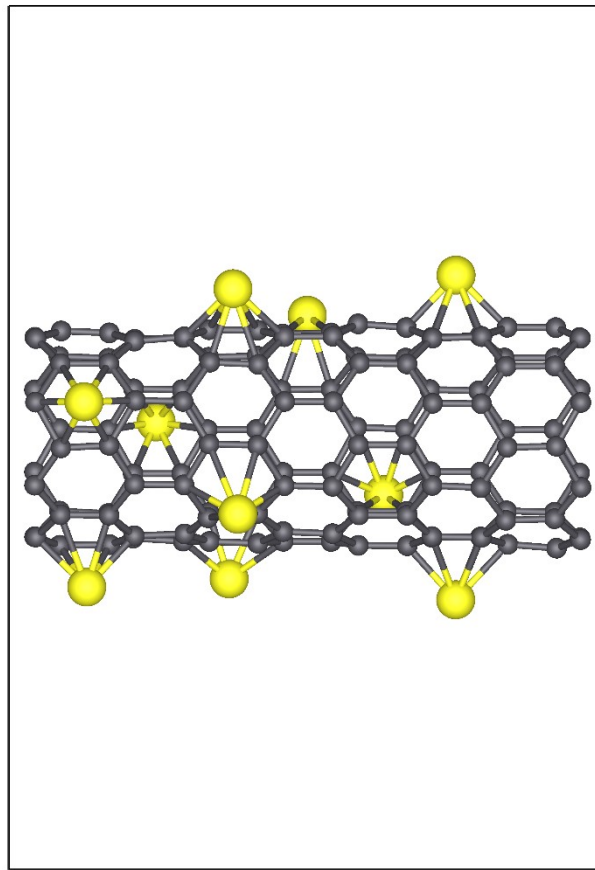
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SWNT(8, 0)+5Li –
SWNT(8, 0) with five
adsorbed Li atoms



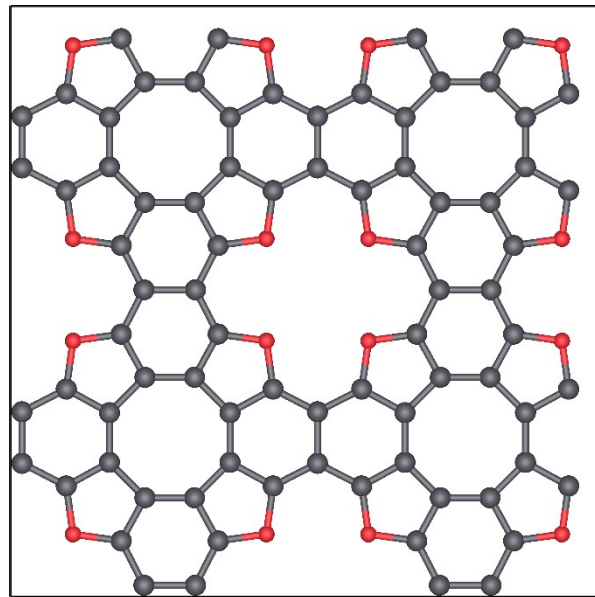
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SWNT(8, 0)+10Li –
SWNT(8, 0) with ten
adsorbed Li atoms



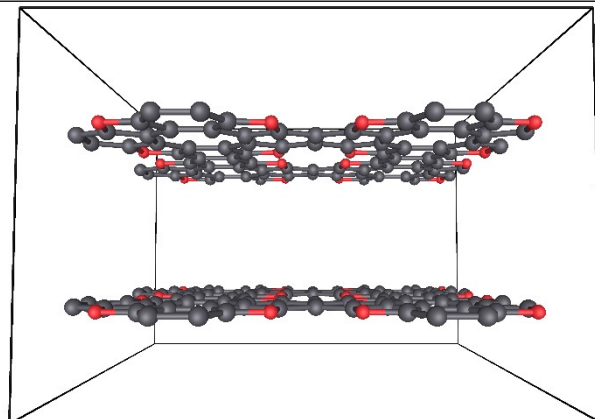
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TOC

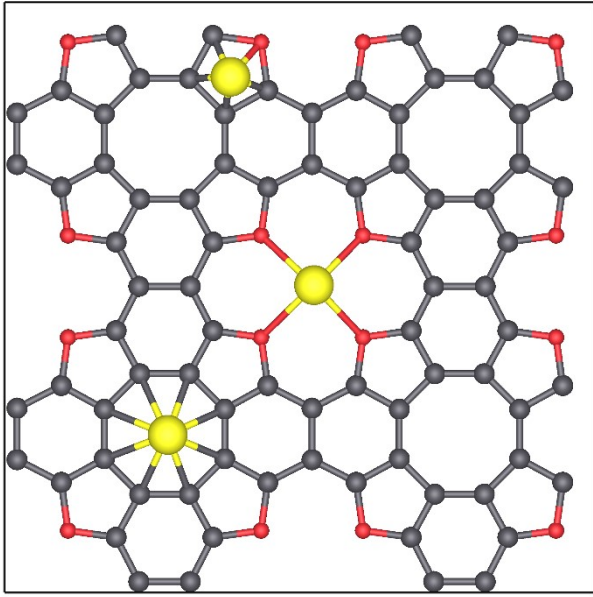
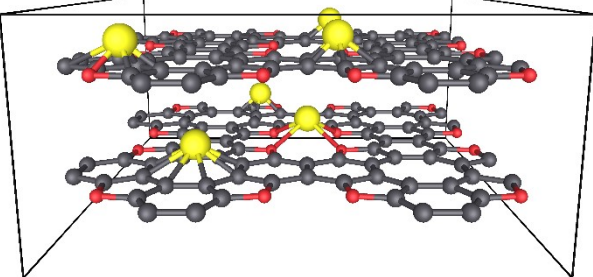
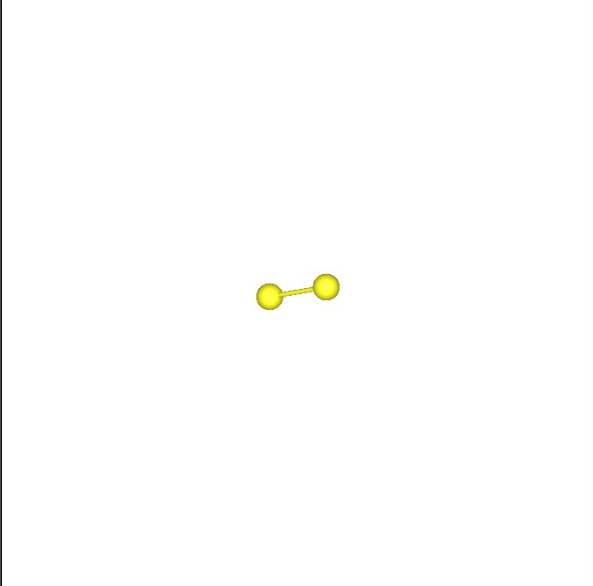


1923

TOC_3D – stacked TOC
structure with three-
dimensional periodicity



479

<p>TOC+3Li – TOC with three adsorbed Li atoms</p>	 <p>A ball-and-stick model of a TOC layer, showing a hexagonal lattice of carbon atoms (grey) and oxygen atoms (red). Three lithium atoms (yellow) are adsorbed on the surface, each coordinated to three oxygen atoms in a trigonal planar arrangement.</p>	<p>951</p>
<p>TOC_3D+3Li – stacked TOC structure with three-dimensional periodicity and three adsorbed 3 Li atoms per TOC layer)</p>	 <p>A 3D model showing multiple stacked TOC layers. Three lithium atoms (yellow) are adsorbed on the surface of one of the layers, coordinated to oxygen atoms (red) in the lattice.</p>	<p>84</p>
<p>Li_2 – dimer</p>	 <p>A ball-and-stick model of a lithium dimer (Li_2), consisting of two lithium atoms (yellow) bonded together.</p>	<p>153</p>


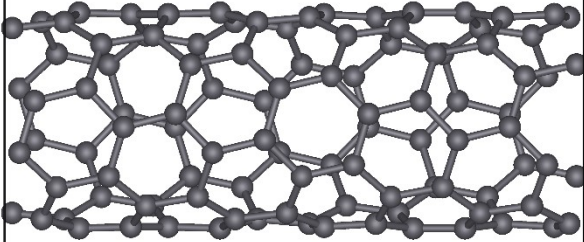
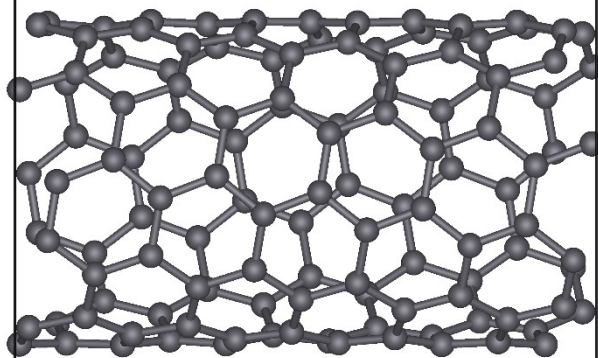
Li ₃ – trimer		140
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Table S2 summarises the external test systems that were not included in the training dataset, together with representative atomic structures and the corresponding simulation cell parameters.

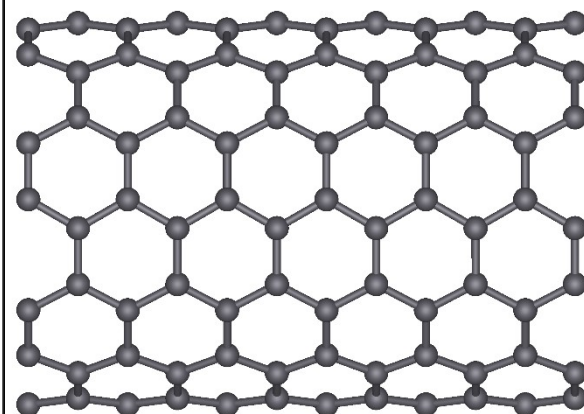
Table S2. Test systems not included in the training dataset

System type	Structure
SWNT(6,2) – single-walled carbon nanotube (6,2), c=15.41	

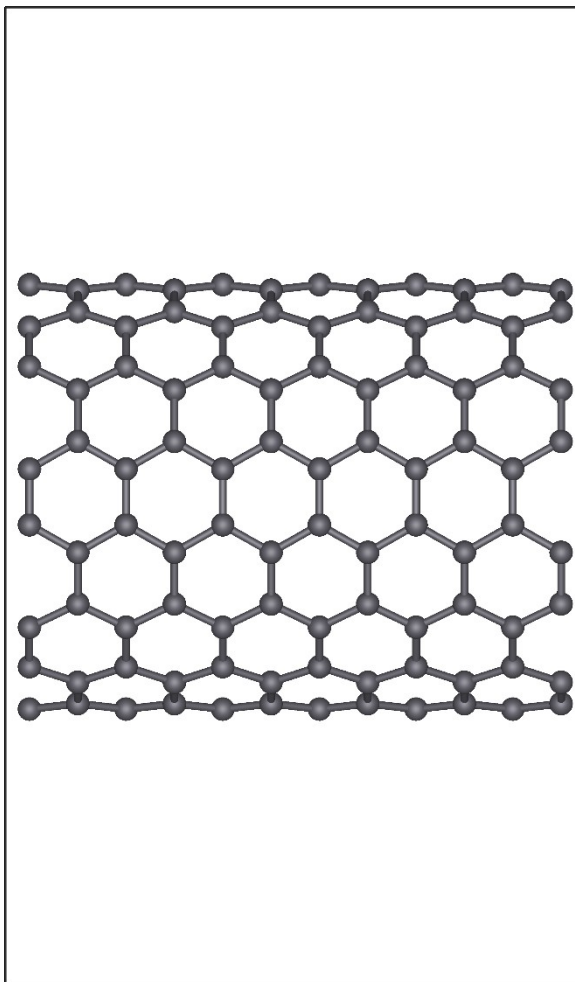
SWNT(7,3) – single-walled
carbon nanotube (7,3),
 $c=13.74$



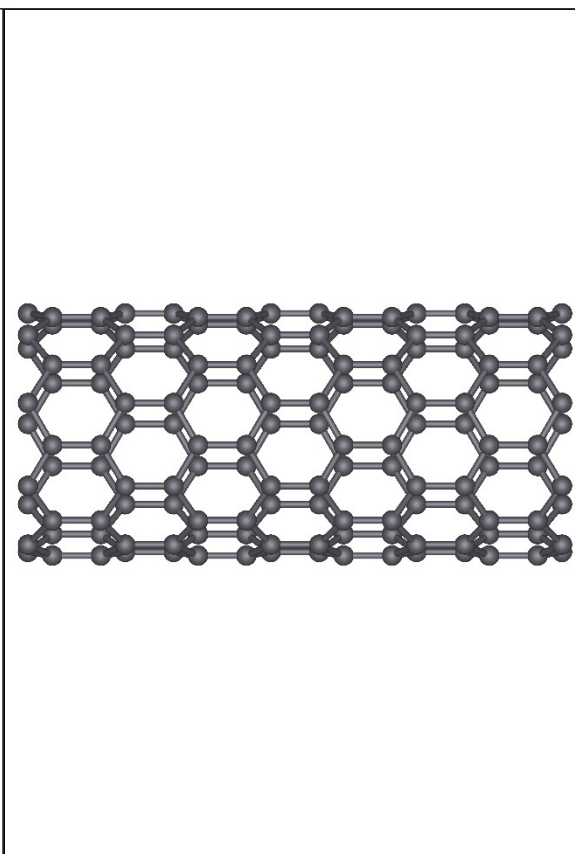
SWNT(7,7) – single-walled
carbon nanotube (7,7),
 $c=14.80$



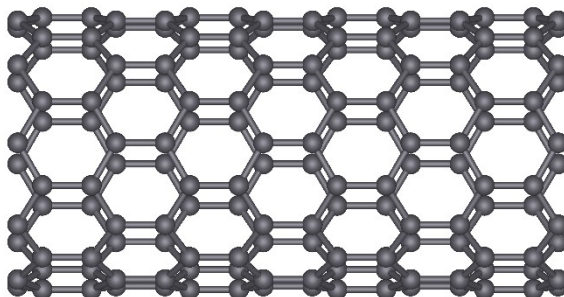
SWNT(8,8) – single-walled
carbon nanotube (8,8),
 $c=14.80$



SWNT(9,0) – single-walled
carbon nanotube (9,0),
 $c=17.10$



SWNT(10,0) – single-walled
carbon nanotube (10,0),
 $c=17.11$



RMSE of energies and force magnitudes

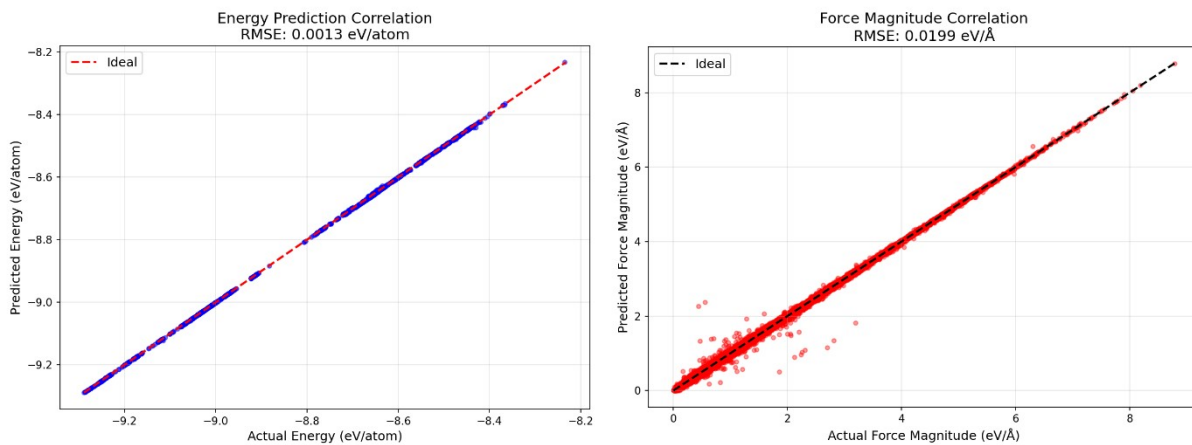


Figure S1. RMSE of energies and force magnitudes for DPA-3.1-fine potential

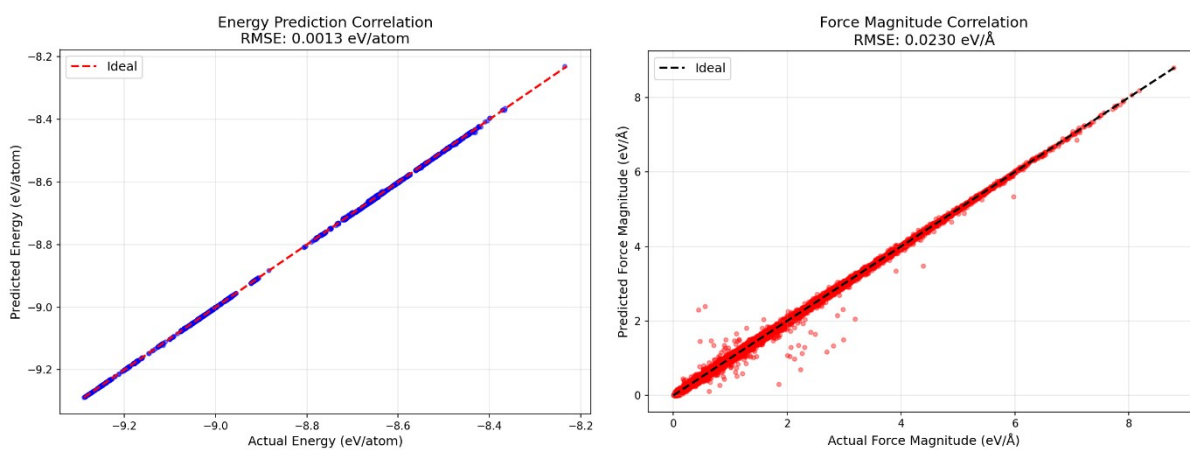


Figure S2. RMSE of energies and force magnitudes for DPA-3.1 potential

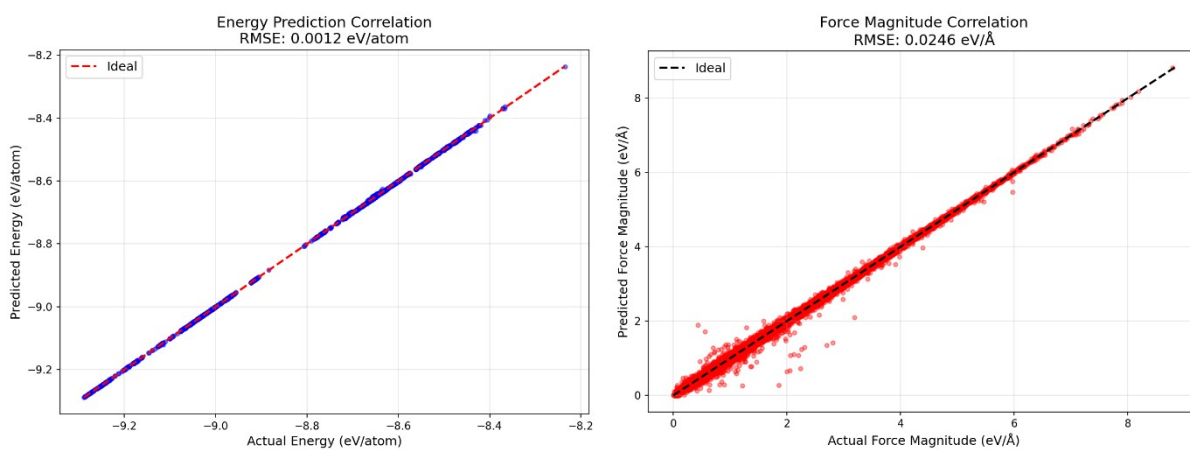


Figure S3. RMSE of energies and force magnitudes for DPA-2.4-fine potential

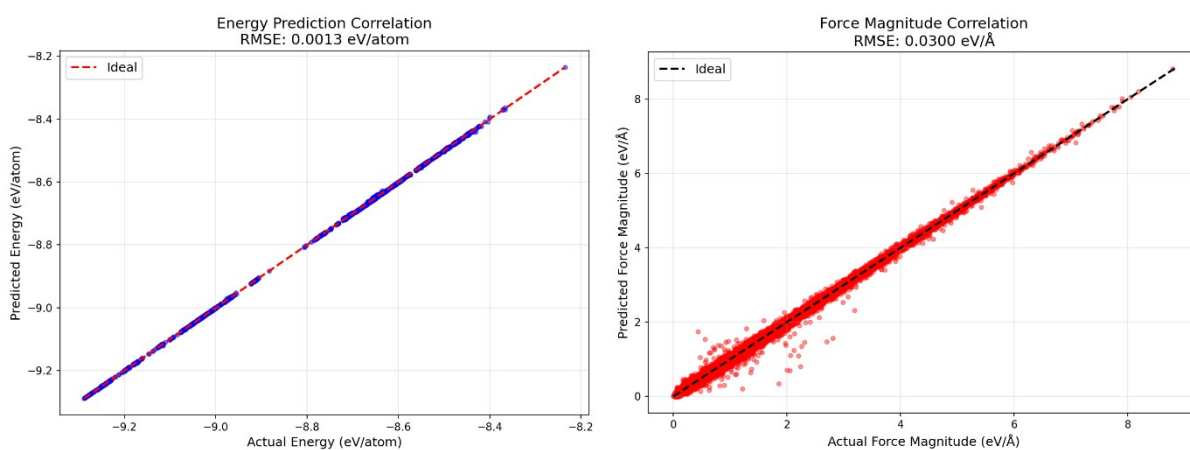


Figure S4. RMSE of energies and force magnitudes for DPA-2.4 potential

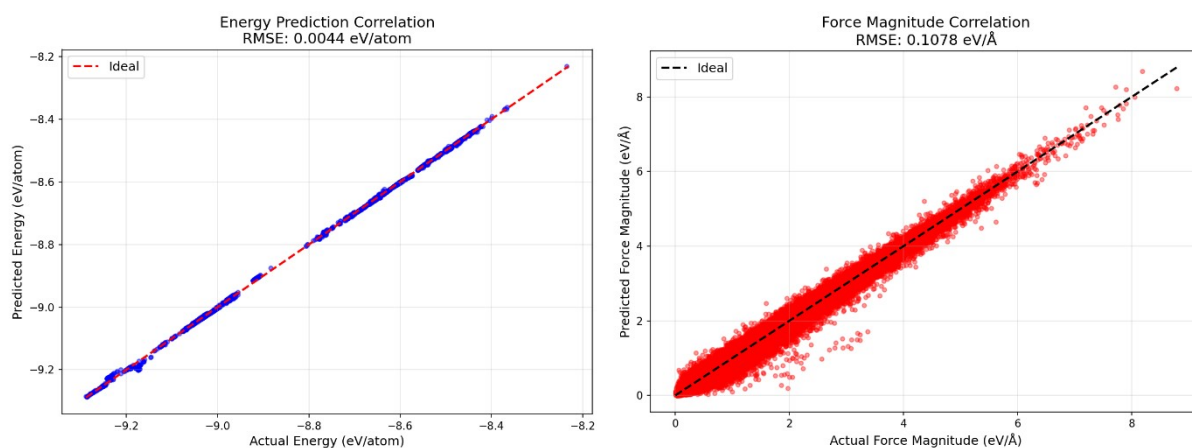


Figure S5. RMSE of energies and force magnitudes for DP-e2 potential

Additional validation results and dataset sensitivity

To evaluate the sensitivity of the potential to the composition of the training dataset, additional models were trained with selected subsets of structures excluded from the database. In particular, two modified datasets were considered: (i) without all iridagraphene-related structures and (ii) without iridagraphene structures lacking adsorbed Li atoms. The resulting fine-tuned potentials are denoted as DPA-3-fine-2 and DPA-3-fine-3, respectively.

Tables S3 and S4 summarize the RMSE values for energies and force magnitudes obtained for different DeePMD models and system classes. In addition to the original models discussed in the main text, these tables also include the modified DPA-3-fine-2 and DPA-3-fine-3 potentials trained using reduced datasets.

Table S3. RMSE of energies (meV/atom) for each model and each trained interatomic potential

model	DPA-2-fine	DPA-2	DPA-3-fine	DPA-3-fine-2	DPA-3-fine-3	DPA-3	DP-e2
graphene	0.4	0.6	0.8	2.2	0.5	0.2	0.4
graphene_vac	1.0	1.6	1.3	0.7	0.6	0.4	26.1
graphene_dvac	0.4	0.2	1.0	5.5	0.4	0.3	11.4
graphite	0.6	0.8	1.4	1.2	0.8	0.4	1.6
SWNT(5, 5)	0.1	0.4	0.6	0.5	0.5	0.3	2.5
SWNT(8, 0)	0.1	0.2	0.6	0.4	0.4	0.3	2.1
biphenilen	0.8	0.2	0.7	0.4	0.5	1.7	6.8
psi-graphene	0.4	0.7	0.5	1.4	0.2	0.6	0.9
irida-graphene	0.6	0.5	1.0	107.5	1.7	2.3	4.4
tpdh-graphene	0.4	0.5	1.2	0.5	0.3	1.6	2.6
N-graphene	0.2	0.6	0.8	1.6	0.6	0.2	14.7
N-pyridinic	0.5	0.5	1.0	1.9	0.6	0.5	12.4
C2N	1.1	1.2	2.0	1.9	0.2	2.1	1.5
C2N_3D	3.3	2.9	2.5	6.1	4.2	2.5	3.1

C2N+3Li	0.9	1.0	1.0	1.8	0.7	1.0	1.4
C2N_3D+3Li	0.0	0.2	0.7	1.3	0.1	0.4	0.2
graphene+1Li	1.3	1.4	1.4	2.2	1.4	1.2	1.4
graphene+3Li	1.0	1.8	0.4	2.5	0.6	0.5	4.8
graphite+3Li	0.6	0.6	0.9	2.2	0.6	0.7	1.4
biphenilen+3Li	0.8	0.7	0.8	0.8	0.7	1.2	4.1
psi-graphene+3Li	0.4	0.5	0.2	2.9	0.1	0.2	7.9
irida-graphene+3Li	0.4	0.4	0.3	126.4	11.8	0.8	2.3
tpdh-graphene+4Li	0.7	0.8	1.0	1.4	0.7	1.1	2.4
SWNT(5, 5)+1Li	0.2	0.6	0.4	1.3	0.4	0.2	2.8
SWNT(5, 5)+5Li	0.7	0.9	0.7	0.7	0.6	0.8	1.3
SWNT(5, 5)+10Li	1.6	1.9	1.1	1.6	1.0	1.0	3.4
SWNT(8, 0)+5Li	0.4	0.7	0.3	0.4	0.1	0.3	4.5
SWNT(8, 0)+10Li	1.4	1.8	0.6	0.8	0.8	0.5	6.0
TOC	0.9	1.2	0.7	0.4	0.2	0.8	1.4
TOC_3D	5.1	4.5	3.6	4.4	3.9	3.5	4.1
TOC+3Li	1.3	1.4	1.4	1.6	1.3	1.4	1.8
TOC_3D+3Li	1.4	1.3	1.9	2.1	1.7	2.1	1.7
Total	1.2	1.3	1.3	26.2	2.4	1.3	4.4

Table S4. RMSE of force magnitudes (meV/Å) for each model and each trained interatomic potential

model	DPA-2-fine	DPA-2	DPA-3-fine	DPA-3-fine-2	DPA-3-fine-3	DPA-3	DP-e2
graphene	13.25	20.75	8.36	19.98	8.56	10.23	86.25
graphene vac	71.88	77.81	59	74.84	57.66	60.54	149.33
graphene_dvac	22.15	27.05	12.06	30.52	11.25	12.91	243.13
graphite	11.1	18.67	10.04	21.24	10.3	14.58	72.31
SWNT(5, 5)	14.32	17.33	13.41	20.46	12.95	14.41	100.15
SWNT(8, 0)	11.14	15.19	10.5	19.36	9.51	11.1	111.88
biphenilen	31.52	43.83	20.48	47.91	20.8	27.66	170.13
psi-graphene	22.61	33.25	15.34	33.31	16.37	19.56	134
irida-graphene	25.2	34.9	19.09	480.69	22.33	23.44	131.61
tpdh-graphene	29.27	42.14	20.43	43.36	19.21	27.41	142.76
N-graphene	7.44	9.38	5.03	13.77	5.12	5.67	73.83
N-pyridinic	58.59	57.92	54.98	59.56	60.37	62.41	152.36
C2N	21.11	29.79	11.41	24.04	12.19	13.03	104.4
C2N_3D	11.05	15.22	11.67	17.85	12.03	10.1	66.06
C2N+3Li	18.02	23.78	12.19	20.39	12.38	13.25	70.76
C2N_3D+3Li	17.95	22.54	10.88	28.8	11.81	12.56	114.42

graphene+1Li	16.89	25.59	11.18	24.27	11.62	13.72	86.1
graphene+3Li	22	30.25	16.22	31.84	16.21	18.09	93.18
graphite+3Li	13	19.3	10.83	21.7	10.98	13.21	61.74
biphenilen+3Li	21.58	27.93	15.85	29.48	16.09	18.98	104.85
psi-graphene+3Li	15.36	21.48	11.56	21.66	11.41	13.06	95.58
irida-graphene+3Li	20.44	26.53	14.51	495.39	150.09	18.14	101.9
tpdh-graphene+4Li	36.67	50.47	26.19	48.95	26.26	34.11	148.77
SWNT(5, 5)+1Li	14.7	21.01	12.92	23.21	12.41	13.83	104.14
SWNT(5, 5)+5Li	33.53	41.04	25.82	44.09	25.78	29.28	122.72
SWNT(5, 5)+10Li	54.35	61.5	42.56	64.6	43.17	45.36	150.14
SWNT(8, 0)+5Li	43.93	51.58	31.47	49.23	34.64	33.44	231.3
SWNT(8, 0)+10Li	66.3	76.48	44.4	72.08	40.08	45.38	214.81
TOC	14.46	23.29	10.08	22.45	10.04	13.4	80.89
TOC_3D	10.84	15.13	12.4	19.39	12.45	14.56	68.62
TOC+3Li	15.95	19.72	12.05	20.33	11.95	13.71	61.98
TOC_3D+3Li	23.32	32.31	18.16	37.18	18.11	24.75	117.03
Total	24.62	29.99	19.86	122.88	39.05	23.01	107.82

Table S5 compares Li adsorption energies predicted by these models against VASP reference data for different classes of carbon-based materials. In the adsorption-site labels, the first number denotes the ring size. For heteroatom-containing rings, the number and type of non-carbon atoms are additionally indicated.

The results show that excluding even relatively small subsets of structurally related configurations may noticeably deteriorate prediction accuracy across multiple material families. This behavior indicates that the model learns transferable structural and chemical relationships shared among different systems rather than merely memorizing local adsorption environments.

Table S5. Li adsorption energies E_{ads} (eV) predicted by different DeePMD potentials trained with modified datasets, compared with VASP reference values

Adsorption center	VASP E_{ads}	DP-3.1-fine		DP-3.1-3M		DP-3.1-fine-2		DP-3.1-fine-3	
		E_{ads}	%	E_{ads}	%			E_{ads}	%
Biphenilen-4	-1.872	-1.879	0.4	-1.535	18.0	-1.848	1.3	-2.104	12.4
Biphenilen-6	-1.926	-1.964	2.0	-1.648	14.4	-1.918	0.4	-2.135	10.9
Biphenilen-8	-1.844	-1.892	2.6	-1.568	15.0	-1.878	1.8	-2.050	11.1
Graphene-6	-1.366	-1.292	5.5	-1.740	27.4	-1.300	4.9	-1.331	2.6
Irida-3	-1.751	-1.774	1.3	-1.682	3.9	-1.320	24.6	-1.536	12.3
Irida-6	-2.025	-2.085	3.0	-1.882	7.0	-1.656	18.2	-1.689	16.6
Irida-8	-2.016	-2.065	2.4	-1.923	4.6	-1.672	17.1	-1.653	18.0
Psigraphene-5	-2.311	-2.623	13.5	-1.793	22.4	-2.362	2.2	-2.807	21.5
Psigraphene-6	-2.276	-2.595	14.0	-1.734	23.8	-2.358	3.6	-2.718	19.4
Psigraphene-7	-2.315	-2.658	14.8	-1.756	24.1	-2.398	3.6	-2.768	19.6
SWNT(5,5)-6	-1.498	-1.417	5.4	-1.764	17.8	-1.330	11.2	-1.452	3.1
Tpdh-4	-2.247	-2.305	2.6	-1.919	14.6	-2.204	1.9	-2.408	7.2
Tpdh-5	-2.298	-2.352	2.4	-2.075	9.7	-2.248	2.2	-2.455	6.8
Tpdh-6	-2.270	-2.336	2.9	-2.055	9.5	-2.220	2.2	-2.435	7.2
Tpdh-10	-2.324	-2.454	5.6	-2.095	9.9	-2.259	2.8	-2.537	9.1
C2N-6-2N	-1.606	-1.616	0.6	-1.742	8.5	-1.622	1.0	-1.720	7.0
C2N-6	-1.396	-1.406	0.7	-1.730	24.0	-1.415	1.4	-1.516	8.6
C2N-18-4N	-4.504	-4.555	1.1	-4.087	9.3	-4.683	4.0	-4.603	2.2
TOC-5-1O	-1.024	-0.950	7.2	-1.220	19.1	-0.732	28.5	-1.043	1.9
TOC-6	-1.185	-1.110	6.3	-1.501	26.7	-0.940	20.6	-1.174	0.9
TOC-8	-1.102	-1.007	8.7	-1.436	30.3	-0.907	17.7	-1.073	2.7

TOC-16-4O	-2.839	-2.803	1.3	-2.812	1.0	-2.596	8.6	-2.792	1.7
C60-5	-1.602	-0.190	88.1	-2.329	45.4	-4.872	204.0	-1.735	8.3
C60-6	-1.599	-0.203	87.3	-2.684	67.9	-4.999	212.6	-1.619	1.2
SWNT(6,2)-6	-1.558	-1.324	15.0	-1.862	19.5	-1.771	13.7	-1.432	8.1
SWNT(7,3)-6	-1.519	-1.494	1.7	-1.765	16.1	-1.302	14.3	-1.550	2.0
SWNT(7,7)-6	-1.520	-1.546	1.7	-1.766	16.2	-1.375	9.6	-1.578	3.8
SWNT(8,8)-6	-1.521	-1.518	0.2	-1.768	16.2	-1.375	9.6	-1.532	0.7
SWNT(9,0)-6	-1.543	-1.534	0.6	-1.790	16.1	-1.356	12.1	-1.617	4.8
SWNT(10,0)-6	-1.345	-1.575	17.1	-1.795	33.5	-1.322	1.7	-1.652	22.8

Table S6 presents the wall-time per molecular dynamics step for tetraoxa[8]circulene calculated using different interatomic potentials on an NVIDIA Tesla V100 (16 GB) GPU. Table S7 summarizes the RMSD values between equilibrium ab initio structures and geometries optimized using the DPA-3-fine potential for various two-dimensional materials and carbon nanotubes.

Table S6. Wall-time per molecular dynamics step (ms) for tetraoxa[8]circulene using different interatomic potentials on an NVIDIA Tesla V100 (16 GB) GPU

Number of atoms	e2_compr	e2	DPA2-compr	DPA2	DPA3
99	0.0091	0.0182	0.026	0.0269	0.0559
396	0.0117	0.0357	0.0672	0.0687	0.1356
792	0.0162	0.0591	0.117	0.12	0.2362
1188	0.0209	0.0827	0.1646	0.1689	
1584	0.0251	0.1057	0.2153	0.2211	
1782	0.0276	0.1176	0.2377	0.2452	
1980	0.0295	0.1282	0.2611	0.2684	
3168	0.0438	0.1991			
4752	0.0622				
6336	0.08				
8019	0.101				
9900	0.1208				
14256	0.1711				
16731	0.2045				
18810	0.2289				

Table S7. RMSD between equilibrium ab initio structures and geometries obtained using the DPA-3-fine potential for various two-dimensional systems and carbon nanotubes

Model	RMSD, pm
TPDH+1Li	0.0063, 0.2947, 0.0016, 0.0163
TOC	0.0003
TOC+1Li	0.0287, 0.3652, 0.0049, 0.0066

C2N	0.0002
C2N+1Li	0.2038, 0.3594, 0.0426
SWNT(5, 5)	0.0001
SWNT(5, 5)+1Li	0.0355
SWNT(6, 2)	0.0007
SWNT(6, 2)+1Li	0.0255
SWNT(7, 3)	0.0004
SWNT(7, 3)+1Li	0.0266
SWNT(7, 7)	0.0004
SWNT(7, 7)+1Li	0.0352
SWNT(8, 8)	0.0004
SWNT(8, 8)+1Li	0.0279
SWNT(9, 0)	0.0004
SWNT(9, 0)+1Li	0.0228
SWNT(10, 0)	0.0009
SWNT(10, 0)+1Li	0.0179

Methodological details of elastic properties determination

The elastic properties of two-dimensional materials and carbon nanotubes were evaluated in the vicinity of the equilibrium configuration using total energy calculations under small uniaxial deformations. All deformations were kept within the linear elastic regime, with strain amplitudes not exceeding a few percent, ensuring a quadratic dependence of the total energy on strain.

For each system, a series of strained configurations was generated by applying longitudinal strain ε to the simulation cell. For two-dimensional materials, the strain was applied in-plane, while for carbon nanotubes it was applied along the tube axis. For each imposed longitudinal strain, the transverse strain component was determined by minimizing the total energy, allowing extraction of the Poisson's ratio.

At each strain value, atomic positions were fully relaxed while keeping the deformed cell parameters fixed. Total energies were calculated both using density functional theory (DFT) and the DPA-3-fine machine-learned potential, ensuring a consistent comparison between the two approaches.

Near equilibrium, the total energy as a function of strain was fitted by a quadratic expression

$$E(\varepsilon) = E_0 + C\varepsilon^2,$$

where E_0 is the energy of the undeformed system and C is the curvature coefficient. Elastic moduli were obtained by normalizing the coefficient C with respect to the effective geometry of the system.

For two-dimensional materials, the two-dimensional Young's modulus was defined as

$$E^{2D} = \frac{2C}{A_0},$$

where A_0 is the equilibrium in-plane area of the simulation cell. When comparison with literature data expressed in three-dimensional units (GPa) was required, an effective layer thickness was introduced. For graphene and related materials, this thickness was taken to be equal to the interlayer spacing in the corresponding bulk material.

For carbon nanotubes, the Young's modulus was calculated by normalizing the curvature coefficient with respect to the effective nanotube volume $V_0 = \pi dtL$,

where d is the nanotube diameter, L is the length of the simulation cell, and $t=0.34$ nm is the effective wall thickness, corresponding to the interlayer spacing in graphite. This definition is consistent with common experimental and theoretical approaches used to determine the elastic moduli of carbon nanotubes [1].

The Poisson's ratio was extracted from the linear relation between transverse and longitudinal strains in the small-deformation limit. All reported elastic constants correspond to zero-temperature values obtained from static total energy calculations.

Methodological details of diffusion barrier determination

To evaluate the diffusion barrier of a lithium atom between neighboring adsorption sites, a series of single-point calculations was performed. The lithium atom was systematically displaced along a straight path connecting the equilibrium positions corresponding to adjacent adsorption sites, while the positions of all other atoms were kept fixed. The calculations were carried out using both VASP and the trained DeePMD potential (DP-3.1-fine). Figures S6–S8 show the atomic structures with indicated adsorption sites for biphenylene, C_2N , and TOC, respectively, along with the corresponding energy profiles for lithium migration between these sites.

For most migration pathways, the DeePMD potential reproduces the DFT energy profiles with good quantitative agreement in the physically relevant energy range corresponding to realistic diffusion barriers. An exception was observed for the TOC 1–2 migration path, where the Li atom approaches an oxygen atom at very short distances. In this case, the DeePMD and DFT energy profiles qualitatively diverge at highly unfavorable configurations corresponding to very large energies (up to ~ 15 eV in the DFT calculations). However, the agreement remains reasonable up to energies of approximately 1.5–2 eV, which are significantly above the thermally accessible diffusion barriers relevant for practical Li migration processes.

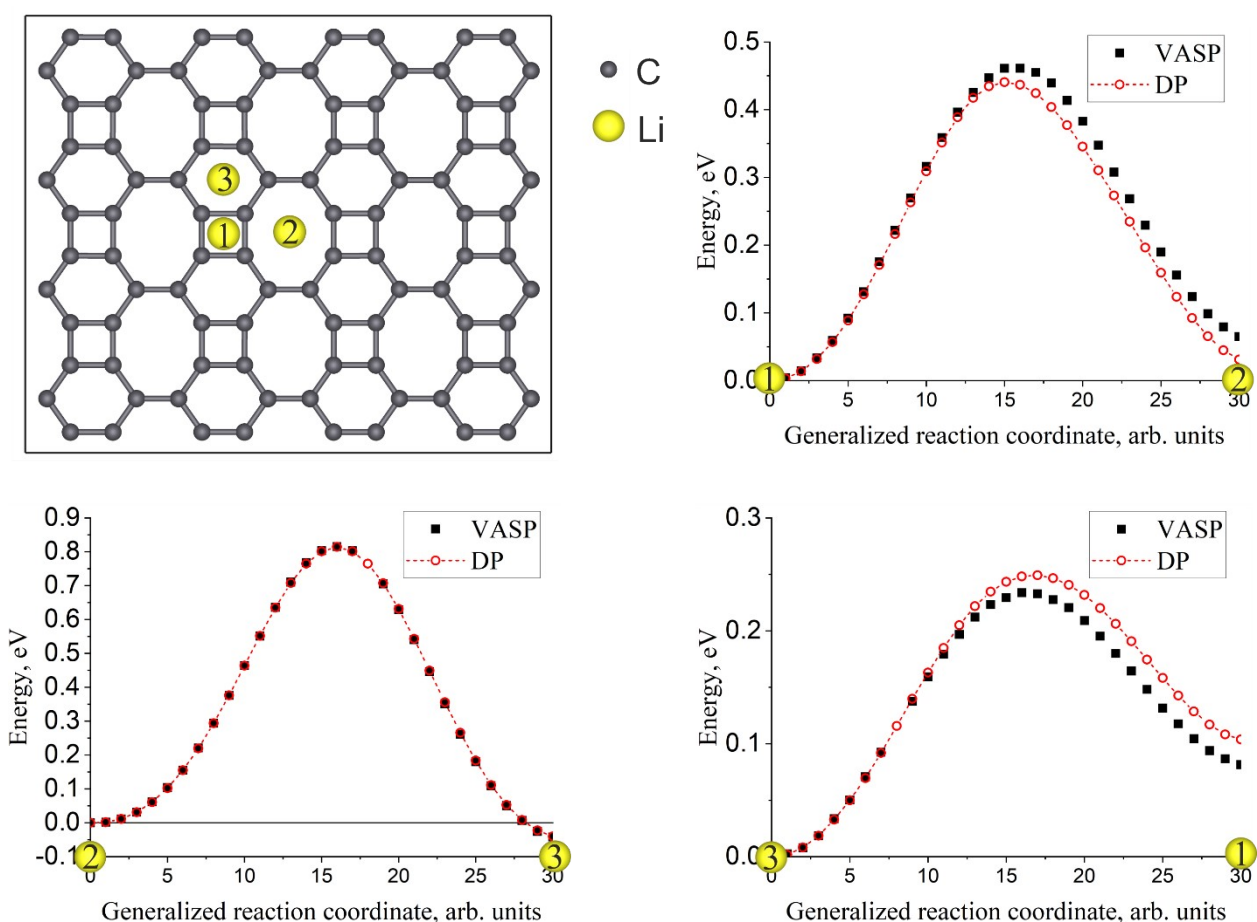


Figure S6. Energy profile for Li atom migration along a linear path between two adsorption sites in biphenylene. Results obtained using VASP and DeePMD are shown for comparison.

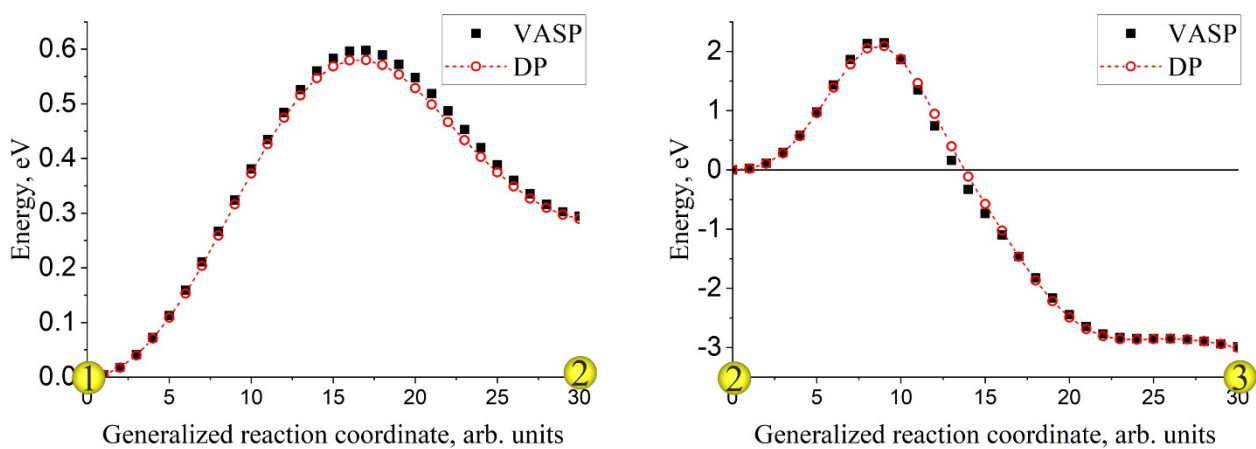
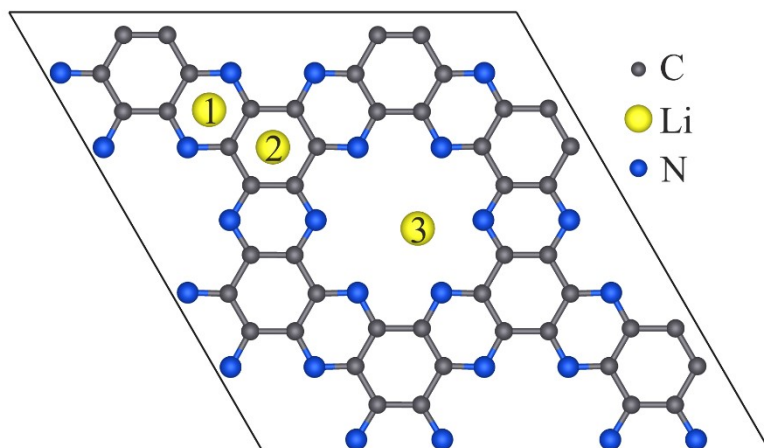


Figure S7. Energy profile for Li atom migration along a linear path between two adsorption sites in C₂N. Results obtained using VASP and DeePMD are shown for comparison.

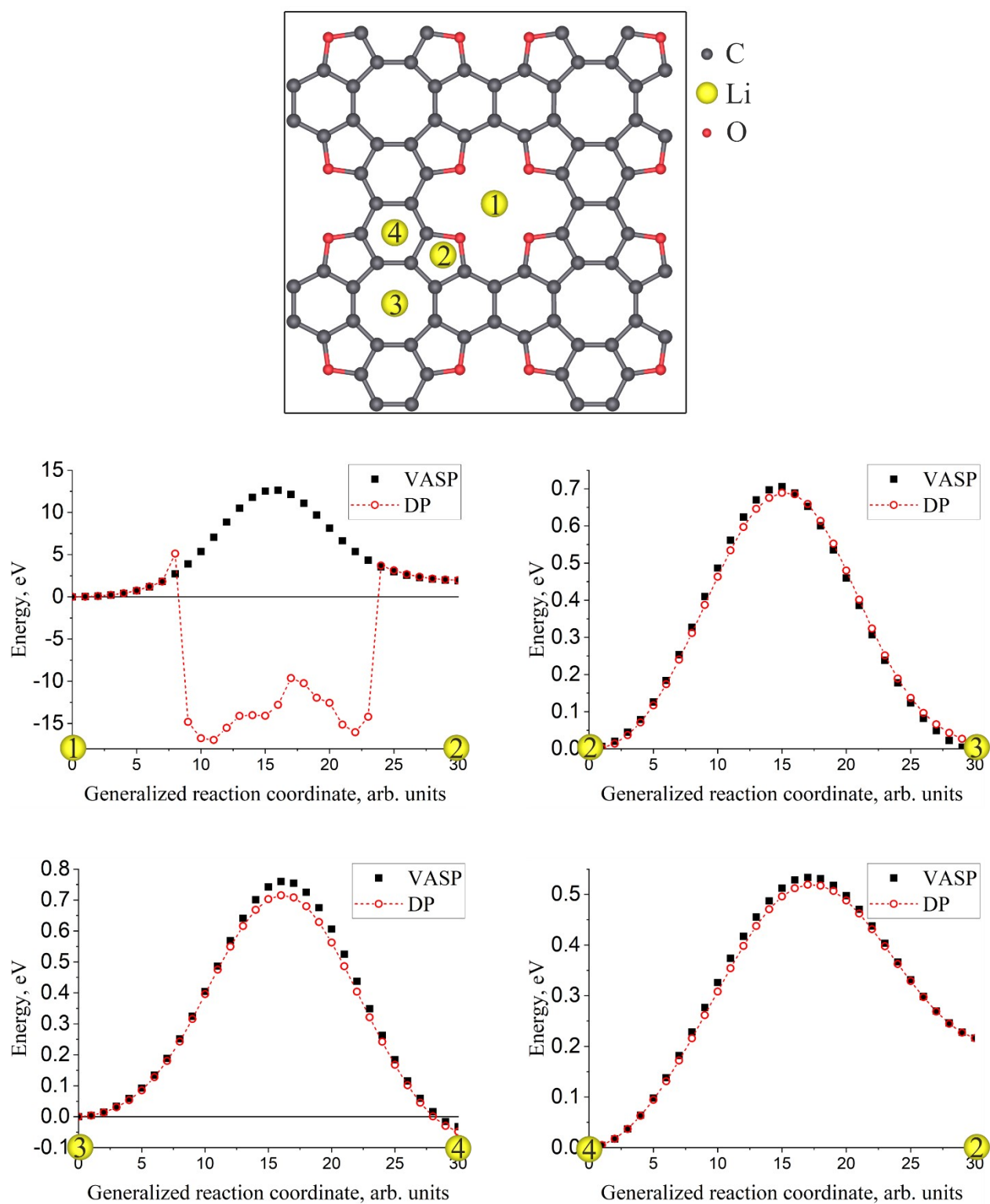


Figure S8. Energy profile for Li atom migration along a linear path between two adsorption sites in TOC. Results obtained using VASP and DeePMD are shown for comparison.

1. Wu Y. и др. Determination of the Young's Modulus of Structurally Defined Carbon Nanotubes // Nano Lett. 2008. T. 8, № 12. С. 4158–4161.