

## Supporting Information

### Searching for Sustainable Refrigerants by Bridging Molecular Modeling with Machine Learning

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**Table S1.** Input dataset for ANN with molar mass and COSMO-RS molecular refrigerants included in this work.

Compound	<i>Mw (g.mol<sup>-1</sup>)</i>	<i>S</i> <sub>2</sub>	<i>S</i> <sub>3</sub>	<i>S</i> <sub>4</sub>	<i>S</i> <sub>5</sub>	<i>S</i> <sub>6</sub>
R41	34.03	0.0000	0.0118	0.0205	0.0053	0.0158
R32	52.02	0.0000	0.0208	0.0061	0.0246	0.0064
R23	70.01	0.0070	0.0062	0.0065	0.0536	0.0000
R161	48.06	0.0000	0.0079	0.0336	0.0112	0.0169
R152a	66.05	0.0000	0.0210	0.0186	0.0219	0.0107
R134a	102.03	0.0022	0.0146	0.0108	0.0599	0.0023
R125	120.02	0.0082	0.0046	0.0155	0.0695	0.0000
R245fa	134.05	0.0039	0.0170	0.0137	0.0741	0.0015
R236fa	152.04	0.0082	0.0074	0.0165	0.0879	0.0000
R227ea	170.03	0.0080	0.0029	0.0254	0.0792	0.0000
R1123	82.02	0.0018	0.0065	0.0233	0.0496	0.0002
R1243zf	96.05	0.0001	0.0186	0.0178	0.0627	0.0000
R1234yf	114.04	0.0003	0.0124	0.0268	0.0636	0.0000
R1234ze(E)	114.04	0.0038	0.0103	0.0216	0.0684	0.0000
R1225ye(Z)	132.03	0.0040	0.0055	0.0290	0.0705	0.0000
R1336mzz(Z)	164.06	0.0062	0.0067	0.0324	0.0860	0.0000
R1233zd(E)	130.49	0.0038	0.0087	0.0296	0.0725	0.0000
R1224yd(Z)	149.49	0.0035	0.0046	0.0375	0.0725	0.0000
R116	138.01	0.0005	0.0117	0.0210	0.0093	0.0000

**Table S2.** Polar soft-SAFT molecular parameters for refrigerants included in this work.

Class	Compound	$m$	$\sigma$ [Å]	$\varepsilon/k_B$ [K]	$\mu \cdot 10^{-30}$ [C·m] <sup>§</sup>	$x_p$	Ref.
<b>Alkane</b>	Ethane	1.392	3.756	202.5	-	-	1
	R41	1.371	3.400	180.3	6.17427	0.50	2
	R32	1.376	3.506	164.5	6.59790	0.75	2
	R23	1.397	3.610	147.9	5.50047	0.90	2
	R161	1.577	3.693	232.3	6.47014	0.33	2
<b>3<sup>rd</sup> gen.</b>	R152a	1.662	3.754	202.3	7.54522	0.50	2
<b>HFCs</b>	R134a	1.813	3.770	169.5	6.86475	0.70	3
	R125	1.887	3.790	165.1	5.21360	0.90	2
	R245fa	2.479	3.675	197.1	5.16690	0.80	2
	R236fa	2.056	4.012	172.4	6.61124	0.90	2
	R227ea	2.131	4.033	190.7	4.85669	1.00	2
	R1123	1.527	3.760	175.3	5.73730	0.80	2
	R1243zf	1.904	3.880	170.0	8.16890	0.50	2
<b>4<sup>th</sup> gen.</b>	R1234yf	1.740	4.082	191.6	6.70790	0.70	3
<b>HFOs</b>	R1234ze(E)	2.044	3.821	204.0	4.80330	0.75	2
	R1225ye(Z)	2.077	3.845	172.4	6.03750	0.80	2
	R1336mzz(Z)	1.806	4.430	195.6	10.6406	0.60	2
<b>4<sup>th</sup> gen.</b>	R1233zd(E)	2.331	3.819	232.6	3.81190	0.80	2
<b>HCFOs</b>	R1224yd(Z)	2.278	3.899	202.4	5.63720	0.85	2

<sup>§</sup>Experimental dipole moment<sup>4-9</sup>

**Table S3.** Output dataset for ANN with polar soft-SAFT molecular parameters for refrigerants included in this work.

Compound	$\mu \cdot 10^{-30}$ [C·m] <sup>§</sup>	$x_p$	$m$	$m\sigma^3$ [Å <sup>3</sup> ]	$m\varepsilon/k_B$ [K]
R41	6.174	0.500	1.371	53.886	247.191
R32	6.598	0.750	1.376	59.300	226.352
R23	5.500	0.900	1.397	65.723	206.616
R161	6.470	0.330	1.577	79.427	366.337
R152a	7.545	0.500	1.662	87.925	336.223
R134a	6.865	0.700	1.813	97.145	307.304
R125	5.214	0.900	1.887	102.728	311.544
R245fa	5.167	0.800	2.479	123.041	488.611
R236fa	6.611	0.900	2.056	132.772	354.454
R227ea	4.857	1.000	2.131	139.787	406.382
R1123	5.737	0.800	1.527	81.171	267.683
R1243zf	8.169	0.500	1.904	111.215	323.680
R1234yf	6.708	0.700	1.740	118.350	333.384
R1234ze(E)	4.803	0.750	2.044	114.028	416.976
R1225ye(Z)	6.038	0.800	2.077	118.066	358.075
R1336mzz(Z)	10.641	0.600	1.806	157.011	353.254
R1233zd(E)	3.812	0.800	2.331	129.835	542.191
R1224yd(Z)	5.637	0.850	2.278	135.025	461.067

**Table S4.** ANN weights and bias for the links between neurons in the input layer, and neurons in the 1<sup>st</sup> hidden layer.

		<i>Neurons in input layer (i)</i>						<i>Neuron bias (b<sub>ij</sub>)</i>
		<i>Mw</i>	<i>S<sub>2</sub></i>	<i>S<sub>3</sub></i>	<i>S<sub>4</sub></i>	<i>S<sub>5</sub></i>	<i>S<sub>6</sub></i>	
<i>Neurons in 1<sup>st</sup> hidden layer (j)</i>	$\mu_1$	0.0199	759.393	-221.629	-53.3194	-28.989	-55.8245	-0.945
	$\mu_2$	0.0599	273.925	29.648	31.886	44.855	-1.577	-10.543
	$\mu_3$	-0.00879	219.773	-202.017	-46.3872	19.301	-10.375	0.764
	$\mu_4$	0.0668	-691.009	-127.769	103.573	3.469	-59.979	-5.633
	$\mu_5$	0.000495	127.437	-3.887	-45.675	-30.769	337.681	1.913
	$\mu_6$	0.00394	800.905	-15.216	93.798	57.655	-72.769	-8.912
	$\mu_7$	-0.0315	521.543	330.952	22.345	1.337	146.780	-3.278
	$\mu_8$	0.0449	-207.964	-48.581	80.760	2.153	-99.433	-2.484
	$\mu_9$	0.0220	538.290	-139.335	-95.831	-40.209	-4.413	-0.301
	$x_{p1}$	0.0437	59.978	-134.377	-172.845	3.124	49.577	-0.0166
	$x_{p2}$	-0.0382	119.168	88.579	139.689	41.318	7.879	-1.839
	$x_{p3}$	-0.0139	-388.942	91.501	-16.887	-26.252	33.729	2.668

**Table S5.** ANN weights and bias for the links between neurons in the 1<sup>st</sup> hidden layer, and neurons in the 2<sup>nd</sup> hidden layer.

		<i>Neurons in 2<sup>nd</sup> hidden layer (k)</i>			
		<i>H<sub>1</sub></i>	<i>H<sub>2</sub></i>	<i>H<sub>3</sub></i>	<i>H<sub>4</sub></i>
<i>Neurons in 1<sup>st</sup> hidden layer (j)</i>	$\mu_1$	-0.0196	0.0292	0.0132	-0.0759
	$\mu_2$	-0.0186	-0.0014	-0.0313	0.0094
	$\mu_3$	0.0869	0.0084	-0.0862	0.0653
	$\mu_4$	-0.0445	0.0294	-0.0162	-0.0631
	$\mu_5$	0.0471	-0.0016	0.0625	-0.0085
	$\mu_6$	-0.0603	-0.0115	0.0047	-0.0019
	$\mu_7$	-0.0109	-0.0379	0.0722	0.0284
	$\mu_8$	-0.1219	-0.0184	-0.0273	-0.0020
	$\mu_9$	-0.0359	-0.0017	0.0258	-0.0207
	$x_{p1}$	0.0020	0.0144	-0.0121	-0.0206
	$x_{p2}$	-0.0001	0.0076	-0.0503	-0.1121
	$x_{p3}$	0.0300	0.0151	0.0457	-0.0402
	<i>Neuron bias (b<sub>jk</sub>)</i>		0.0973	-0.0019	0.0408

**Table S6.** ANN weights and bias for the links between neurons in the output layer, and neurons in the 1<sup>st</sup> or 2<sup>nd</sup> hidden layers.

		<i>Neurons in output layer (O)</i>				
		$\mu \cdot 10^{-30}$	$x_p$	$m$	$m\sigma^3$	$m\epsilon/k_B$
<i>Neurons in 1<sup>st</sup> hidden layer (j)</i>	$\mu_1$	11.898	-	-	-	-
	$\mu_2$	-0.114	-	-	-	-
	$\mu_3$	-8.478	-	-	-	-
	$\mu_4$	-1.612	-	-	-	-
	$\mu_5$	-1.039	-	-	-	-
	$\mu_6$	-3.692	-	-	-	-
	$\mu_7$	-0.111	-	-	-	-
	$\mu_8$	1.618	-	-	-	-
	$\mu_9$	4.457	-	-	-	-
	$x_{p1}$	-	-0.524	-	-	-
	$x_{p2}$	-	-0.804	-	-	-
	$x_{p3}$	-	-0.482	-	-	-
	<i>Neurons in 2<sup>nd</sup> hidden layer (k)</i>	$H_1$	-	-	-100.481	1358.022
$H_1$		-	-	411.074	-8831.445	178751.003
$H_1$		-	-	119.880	-2067.174	51621.750
$H_1$		-	-	226.605	-4999.385	96884.334
<i>Neuron bias (b<sub>io</sub>)</i>		2.699	0.889	2.073	102.779	430.899

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