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Multi-objective optimization and Coevolutionary search for materials with optimal properties

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And is based on the following papers:

- Z. Allahyari, A. R. Oganov. Nonempirical Definition of the Mendeleev Numbers: Organizing the Chemical Space. *J. Phys. Chem. C*,124, 43, 23867–23878 (2020).
- Z. Allahyari, A. R. Oganov. Coevolutionary search for optimal materials in the space of all possible compounds. *NPJ Comput Mater*, 6, 55 (2020).
- Z. Allahyari, A. R. Oganov. Multi-objective Optimization as a Tool for Material Design. In: Andreoni W., Yip S. (eds) Handbook of Materials Modeling. Springer, Cham (2020).





Objectives

- 1. To solve the challenging and the main problem of material science: predicting optimal materials in any desired properties, let's say in the space of all binary systems.
- 2. To make sure, that the predicted materials have a high chance to be synthesized. We don't need unrealistic materials which are only perfect in computers.
- 3. Such searches are huge. To reduce the time and computational cost of the searches, without significant loss in the results.
- 4. To have an automatic and unbiased search for hard/superhard materials. Is diamond the hardest material in nature? Are there any new super(hard) materials?





Outline

Search for materials with optimal target properties:

Mendelevian Serach-MendS (algorithm, method)[1][2].

Defining a good chemical space.

Multi-objective Pareto technique.

Results on searching (using MendS) for hard/superhard and low energy binary materials under zero pressure in the entire chemical space.

[1] Patent No: WO2018009090A1

[2] NPJ Comput. Mater. 6, 55 (2020).





What is coevolution and the difference of evolutionary algorithm and coevolutionary algorithm?



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Figure 1: Cooperative coevolutionary architecture from the perspective of species number one.



Child making in evolutionary algorithm is so obvious, but what about coevolutionary algorithm?



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What is the child of

C-Fe & P-W

Chemical scale suggested by Pettifor (1984)



• Solid. State. Commun., 51, 31-34 (1984).

Journal of Physics C: Solid State Physics, 19, 3 (1986).

• New J. Phys., **18**, 093011 (2016).

• J. Alloys. Compd., 317, 26-38 (2001).



- 1929 Goldschmidt's law of crystal chemistry: the crystal structure is determined by stoichiometry, **atomic size**, and **polarizability** of atoms/ions [1].
- 1932 Linus Pauling introduced the concept of **electronegativity** [2].
- 1955 Ringwood modification: electronegativity is as another important parameter for determining the crystal structure [3].
- 1990 Nagle: **electronegativity** and **polarizability** are strongly correlated [4].

$$\chi_{\alpha} = 1.66 (n/\alpha)^{1/3} + 0.37$$

 χ_{α} = electronegativity, α = polarizability & *n* = valence.

[1] Trans. Faraday Soc. 25, 253 (1929)

[2] J. Am. Chem. Soc. 54, 3570–3582 (1932)



Our redefined MN, and it's comparison with Skoltech the Pettifor's MN.



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Pettifor maps of Lyakhov-Oganov model of hardness for different MNs



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Chem. Met. Alloy. 1, 1-23 (2008).
 J. Phys. C Solid State Phys. 19, 285 – 313 (1986).

[3] New J. Phys. 18, 093011 (2016).
[4] J. Phys. Chem. C, 124, 43, 23867–23878 (2020).

Clustering using density peaks



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Science, **344**, 1492 – 1496 (2014).

Local density(
$$\rho_i$$
) $\rho_i = \sum_j \chi(dij - dc)$
 $d_c = \text{cutoff distance}$
 $\chi(x) = 0 \quad if \quad x < 0$
 $\chi(x) = 1 \quad if \quad x > 0$

Cluster centers are points with highest local density.

 d_p = property difference cutoff between a cluster member and the cluster center. Ideal MN = an imaginary MN, that covers all systems in minimum number of clusters (N_{min}).



Evaluation of the MNs using clustering method

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Number of clusters

AN: atomic number; PN: Periodic number; MN_P: Pettifor's MN; MN_m: modified MN; USE: MN in this work.

Skoltech MendS algorithm and its important Variation operators Skolkovo Institute of Science and Technology **Reactive heredity operator:** Finding the child of two parent red and green using A-B & C-D → A-C , A-D, B-C or B-D chemical heredity operator. 1st parent Child Defining the Chemical space 2nd parent Selection Electronegativity probability EA EA EΔ EA Atomic Radius Ω **Evaluate EA results** Using chemical mutation operator we mutate an atom to find Ranking the structures using MO Pareto technique (Ranking Parent is assigned as a Fitness) End Number of generation = N No Selection of adequate systems for heredity and mutation 0.5 Atomic Radius Random selection Child

EA

Implementation of multi-objective Pareto

method and its test on Mo_xN_y



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J. Phys. Chem. C, 120 (20), pp 11060-11067 (2016).





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Efficiency of the algorithm in the system selection – improving the Pareto front

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	Compounds	H_v (GPa)	$K_{1C}(\mathrm{MPa.m^{1/2}})$	Instability (eV/atom)	Space group		Compounds	H_v (GPa)	$K_{1C}(\mathrm{MPa.m^{1/2}})$	Instability (eV/atom)	Space group
Carbon	С	92.7	6.33	0.13	Fd3m	Boron	В	38.9	2.87	0	$R\bar{3}m$
	С	93.6	6.36	0.139	$P6_3/mmc$		В	44.8	3.29	0.136	$Cmc2_1$
B-S	B_4S_3	30.5	1.83	0.102	Cmcm	B-N	BN	$63.4,(62.8)^{25},[46-80]^{20}$	5.1	0.075	$F\bar{4}3m$
	MoB_2	28.5,(33.1) ⁴⁴ ,[24.2] ⁵³	3.76	0	R3m		TcB	31,(30.3)54	3.83	0.013	$P\bar{3}m1$
	MoB ₃	35.3	3.74	0.035	$P\bar{3}m1$		TcB_3	27.2,(29)55	3.6	0	$P\bar{6}m2$
	MoB ₃	32.2	3.63	0.077	A2/m		TcB ₃	33.1	3.79	0.003	$P\bar{3}m1$
		35.3,(37.3)44	3.63	0.017	$P6_3/mmc$		TcB_4	31.8	3.56	0.069	$P2_1/m$
		33.1,(31.8) ⁴³	3.57	0.011	R3m		TcB ₄	30.2	3.54	0.069	R3m
Mo-B	MoB ₄	35.4	3.57	0.099	Pmmn	Тс-В		30,(32)55	3.57	0.027	$P6_3/mmc$
	MoB_5	35.7	3.62	0.054	$P\bar{6}m2$		TcB7	35.9	3.35	0.084	R3m
	MoB_8	36.6	3.24	0.118	R3m		TcB_8	33.9	3.3	0.113	R3m
	Mo_2B_3	32.2	3.95	0.029	Imm2		Tc_3B_5	30.6	3.87	0	$P\bar{6}m2$
	Mo_2B_3	30.4	3.87	0.043	Cmcm						
Si-C	SiC	33.3,(33.1) ³¹ ,[28] ³¹	2.94	0	$F\bar{4}3m$	B-P	BP	37.2,(29.3) ³¹ ,[33] ³¹	2.46	0	$F\bar{4}3m$
	SiC	33.1	2.94	0.001	R3m		B_6P	41.1	2.87	0	R3m
	VB	39.1,(38.3) ⁴⁹	3.66	0	Cmcm		MnH	29.5	3.2	0	$P6_3/mmc$
	VB_2	37.3,(39.5) ⁴⁹ ,[27.2] ⁵⁰	3.75	0	P6/mmm		MnH	27.9	3.14	0.013	$R\bar{3}m$
	VB ₅	40	3.36	0.158	$P\bar{6}m2$		MnH	26.3	3.07	0.044	$Fm\bar{3}m$
V-B	VB_7	39.7	3.19	0.143	P3m1	Mn-H	Mn_3H_2	26.8	3.22	0.017	R32
	VB_{12}	44.5	3.34	0.125	I4/mmm		Mn_3H_2	27	3.26	0.019	$P6_3/mcm$
	V_3B_4	37.8	3.74	0	$P\bar{4}m2$		Mn_4H_3	27.6	3.23	0.002	P2/m
	V_3B_4	35.9,(38.2) ⁴⁹	3.7	0.006	Immm		Mn_6H_5	27.3	3.17	0.011	A2/m
	(MnB ₃)	32.2	3.5	0.029	$P\bar{6}m2$		FeB3	30.2	3.32	0	$P2_1/m$
	MnB_4^{\dagger}	40.7	3.65	0.009	Pnnm		FeB_4	35.7	3.06	0.021	Immm
Mn-B	MnB_4	38.2	3.56	0.1	R3m	Fe-B	FeB_4 ‡	32	3.31	0.039	$R\bar{3}m$
		38.1,(40.5) ⁵¹ ,[37.4] ⁵²	3.76	0	$P2_1/c$		FeB_4	42.7	3.31	0.063	A2/m
	MnB ₅	32.7	3.38	0.097	$P\bar{6}m2$			28.6,(24.4) ⁶¹ [62] ⁵⁹	3.32	0.002	Pnnm
	MnB ₁₃	40.4	2.9	0.181	Pm		$\left(Fe_2B_{11} \right)$	33.8	3.37	0.081	Pm

Ashby plot of hardness vs. toughness for predicted phases



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J. Phys. Chem. Lett., 8 (4), pp 755–764 (2017). Cr-N, Cr-B

J. Phys. Chem. Lett., 9 (12), pp 3470-3477 (2018). W-B





Tungsten carbide WC Hardness 30 GPa







Conclusion

- A well-defined chemical space can be obtained using the most significant properties of elements (electronegativity and atomic size and binary systems with similar properties are nearby in this space.
- Multi-objective Pareto technique makes it possible to search for materials optimal in more than one property, and works efficiently at least for two properties.
- Combination of coevolutionary algorithm, MO Pareto technique and well-structured chemical space - MendS – works efficiently in searching for materials optimal in multiple target properties.
- Our Mendelevian search suggests that diamond is the hardest material (among binaries) in nature harder cannot be found.
- MendS works at arbitrary pressure and for binary, ternary systems.



Thank you for your attention!

Any questions?

