

Supplemental Material for *Efficient first principles based modeling via machine learning: from simple representations to high entropy materials*

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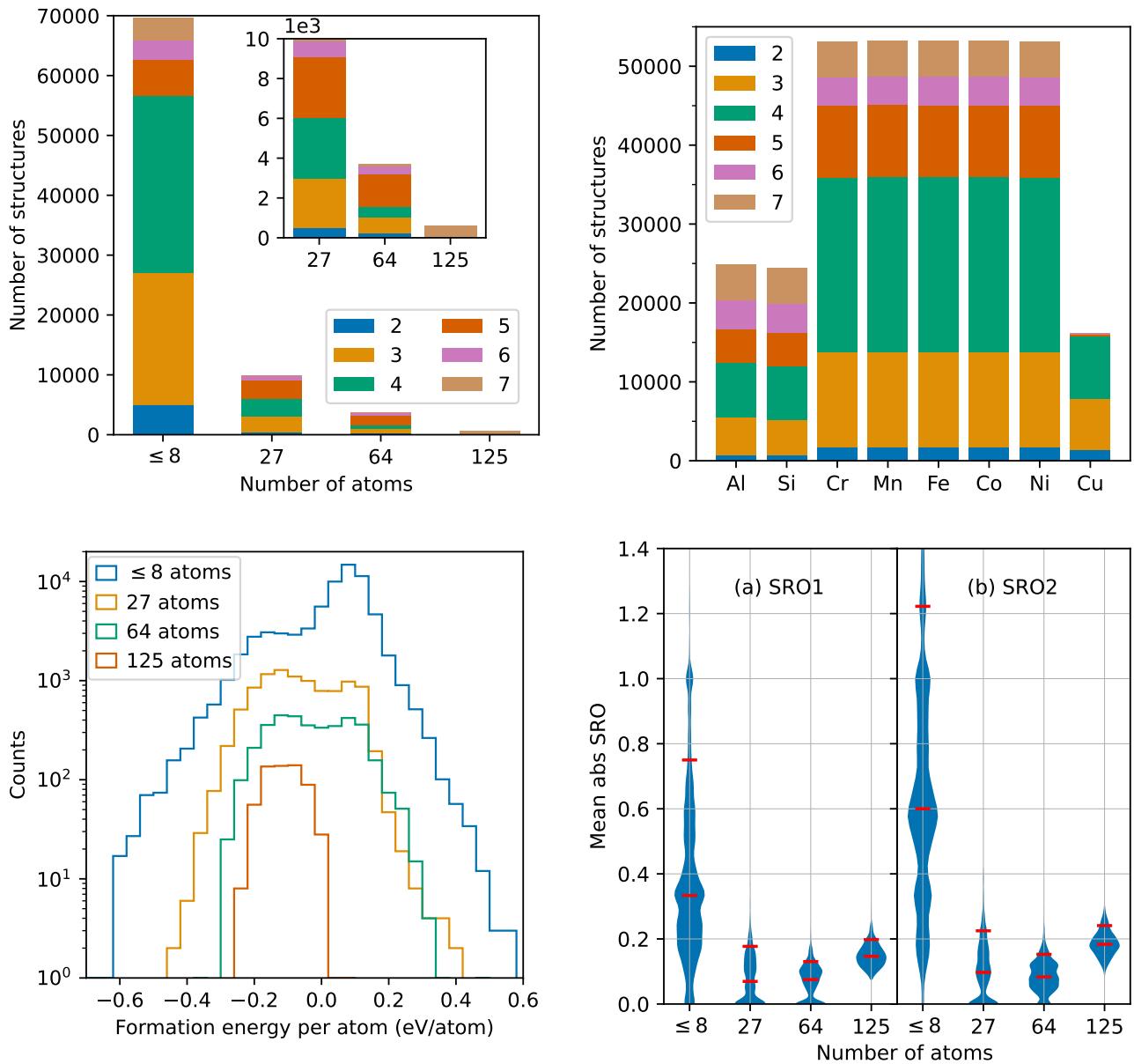
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Supplementary Table I. Number of structures as functions of number of atoms and number of elements.

# atoms	# elements					
	2	3	4	5	6	7
2	40	0	0	0	0	0
3	252	210	0	0	0	0
4	797	4342	2633	0	0	0
5	745	2643	2669	108	0	0
6	2707	11933	18402	3000	0	0
7	0	0	0	0	0	854
8	434	2970	5790	3049	3132	2865
27	486	2505	3021	3077	783	56
64	229	797	521	1641	400	108
125	0	0	0	0	0	598

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Supplementary Figure 1. Distribution of structures as functions of system size (upper left), element (upper right), formation energy (lower left), and SROs (lower right). Upper left panel: The legend indicates the number of components, and the inset is an enlarged view for SQSs. Upper right panel: The legend indicates the number of components, and the X axis indicates the element contained in the structures. Lower right panel: The violin plot indicates the distribution of structures as functions of SRO and system size for the first and second coordination shells.

Supplementary Table II. Hyperparameter search results. We perform a hyperparameter grid search based on 5-fold cross validations within the in-distribution training set. The ranges of hyperparameter values, the best values, and the values used in the paper are reported. The comment row mentions the best MAE and the current MAE, which are practically the same.

Model	Hyperparameters	Value range	Best	Used in paper
Random forest	bootstrap	[True, False]	False	False
	max_depth	[5, 10, 15, 20, None]	None	None
	max_features	[0.1, 0.2, 0.3, 0.45]	0.3	0.3
	n_estimators	[50, 100, 150, 200]	200	100
Comment	Best vs. current MAE: 15.9 vs. 16.0 meV/atom			
XGBoost	n_estimators	[500, 1000, 2000, 3000]	2000	500
	learning_rate	[0.1, 0.2, 0.3, 0.4]	0.2	0.4
	colsample_bytree	[0.3, 0.5, 0.7, 0.9]	0.9	0.5
	colsample_bylevel	[0.3, 0.5, 0.7, 0.9]	0.3	0.7
	num_parallel_tree	[4, 6, 8, 10]	10	6
Comment	Best vs. current MAE: 13.6 vs. 14.4 meV/atom			