Electronic Supplementary Information

Interpretable Machine Learning-Assisted

Screening of Perovskite Oxides

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Fig. S1 Pairwise Pearson correlation coefficient heatmap of the 34 features screened from the 291 features.



Fig. S2 Confusion matrices of the (a) ABC-23, (b) GBC-23, (c) LRC-23, and (d) RFC-23 models.



Fig. S3 ROC curves of the (a) ABC-23, (b) GBC-23, (c) LRC-23, and (d) RFC-23 models.



Fig. S4 The (a) confusion matrix and (b) ROC curve of the XGBC-ref model.



Fig. S5 Performance of the RFR-144, GBR-144, and ABR-144 regression models.



Fig. S6 Performance of the XGBR-ref model that was trained on the dataset of ref 22.

Table S1 The 23 optimal features and the 11 removed features during feature selection for classification model training.

	BL_AB_ratio, IR_AB_ratio, B2_HOMO, B1_LUMO, EN_B+,				
23 optimal features	EN_AB_avg , $B1_ZR$, $B1_ME$, ME_B -, ME_AB_avg ,				
	ME_AB_ratio, IV_AB_avg, IV_AB_diff, IV_AB_ratio,				
	CR_AB_diff, AR_AB_diff, AV_B+, MN_B+, FIP_AB_ratio,				
	SHC_AB_diff, EC_B+, EC_AB_diff, EC_AB_ratio				
11 removed features	$IR_AB_diff, \Delta EN_{BO}*OF, ME_B^+, ME_AB_diff, B1_IV,$				
	CR_AB_ratio, B1_AR, AR_AB_ratio, AV_AB_avg, AV_AB_diff,				
	AV_AB_ratio				

Model	Hyperparameter	Model	Hyperparameter
ABC-23	'learning_rate': 0.5,	ABR-144	'learning_rate': 0.2
	'n_estimators': 250		'n_estimators': 100
GBC-23	'max_depth': 6,	GBR-144	'max_depth': 5
	'n_estimators': 250		'n_estimators': 150
LRC-23	'C': 100,	RFR-144	'max_depth': 13
	'penalty': '12'		'n_estimators': 200
RFC-23	'max_depth': 13,	XGBR-144	'max_depth': 6
	'n_estimators': 100		'n_estimators': 100
XGBC-23	'max_depth': 13,		
	'n_estimators': 100		

Table S2 Hyperparameters for various machine learning models.

Formula	E _h -DFT	E _h -predicted	Formula	<i>E_h</i> -DFT	E _h -predicted
	(meV atom ⁻¹)	(meV atom ⁻¹)		(meV atom ⁻¹)	(meV atom ⁻¹)
Ba ₂ CaOsO ₆ ¹	0	9	BaPdO ₃	29	1
Ba ₂ CeZrO ₆ ²	24	19	CaFeO ₃ ³	52	0
Ba ₂ CePtO ₆	0	0	EuHfO ₃ ⁴	0	6
Ba ₂ CeHfO ₆	23	18	EuNbO ₃ ⁵	24	6
Ba ₂ DyNbO ₆	0	2	EuVO ₃ ⁶	43	0
Ba ₂ ErNbO ₆ ⁷	0	8	PrCuO ₃ ^{8,9}	25	7
Ba ₂ EuReO ₆ ¹⁰	0	5	SrFeO ₃ ¹¹	0	0
Ba ₂ HoMoO ₆	0	0	YbTaO ₃ ¹²	26	0
Ba ₂ HoNbO ₆ ¹³	0	0	BaTaO ₃ ¹⁴	32	6
Ba ₂ LuMoO ₆ ¹⁵	0	7	KReO ₃ ¹⁶	37	14
Ba ₂ LuNbO ₆ ¹⁷	0	0	NaReO ₃ ¹⁶	42	11
Ba ₂ LuRuO ₆ ¹⁸	0	18	SrNiO ₃ ¹⁹	46	0
Ba ₂ LuSbO ₆ ²⁰	0	1	Ba ₂ BiLaO ₆ ²¹	28	30
Ba ₂ LuTaO ₆ ²²	0	0	Ba ₂ BiDyO ₆ ²¹	3	36
Ba ₂ NbVO ₆ ²³	0	5	Ba ₂ BiCeO ₆ ²¹	48	38
Ba ₂ NbFeO ₆	0	0	Ba ₂ BiSmO ₆ ²¹	8	35
Ba ₂ NbInO ₆	0	3	Ba ₂ BiGdO ₆ ²¹	6	26
Ba ₂ SmMnO ₆	45	26	Ba ₂ BiTbO ₆	11	37
Ba ₂ TaTlO ₆	0	22	Ba_2BiLuO_6	2	32

Table S3 Comparison of the E_h values predicted by the XGBR-144 model and calculated by density functional theory that are not included in the input dataset.

Ba ₂ TmNbO ₆	0	25	Ba ₂ BiYO ₆	3	36
Ba ₂ TmRuO ₆	0	33	Eu ₂ CrSbO ₆	39	7
Ba ₂ TmMoO ₆	0	30	Eu ₂ CrSnO ₆	47	12
Ba ₂ ZrSnO ₆	0	20	Eu ₂ FeWO ₆	29	3
Ba ₂ ZrTiO ₆	5	19	Eu ₂ HfTiO ₆	17	6
Ba ₂ LaFeO ₆	42	0	Eu ₂ HfSnO ₆	22	5
BaSrMo ₂ O ₆	47	1	Eu ₂ HfFeO ₆	44	17
BaSrCoWO ₆ ²⁴	15	0	Eu ₂ LuTaO ₆	22	6
Sr2HfSnO ₆	19	18	Eu ₂ MgWO ₆	0	6
Sr2HfTiO ₆	22	12	Eu ₂ MnWO ₆	0	0
Sr2HfFeO ₆	42	5	Eu ₂ NbFeO ₆	37	1
Sr2HfCrO ₆	33	6	Eu ₂ NiWO ₆	34	3
Sr2HfZrO ₆	44	15	Eu ₂ TaAlO ₆	0	10
Sr ₂ NiRuO ₆ ²⁵	0	2	Eu ₂ TiNbO ₆	8	6
Sr ₂ VWO ₆ ²⁶	13	0	Eu ₂ TmTaO ₆	36	17
Sr ₂ FeCoO ₆ ²⁷	20	0	Eu ₂ VWO ₆	33	0
Sr ₂ FeHfO ₆ ²⁸	42	5	Eu ₂ ZnWO ₆	0	4

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