

Supplementary Materials

Data-driven design of high-abundance rare earth permanent magnets for low-carbon and resource-efficient production

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Supplementary Table 1. Essential parameters and settings used in the MLP regression model

Parameter	Setting
Model	MLP regression
Hidden layer structure	(128, 64, 32)
Activation function	relu
Optimizer	Adam
Maximum iterations	3000
Random seed	42
Data scaling	Min–Max normalization
Early stopping	Enabled
Validation fraction	0.15
L2 regularization coefficient (α)	1.048×10^{-6}
Batch size	16
Learning rate	9.65×10^{-3}
Hyperparameter optimization method	Bayesian optimization
Bayesian optimization iterations	30
Optimized hyperparameters	α , learning rate, batch size
α search range	$1 \times 10^{-6} - 1 \times 10^{-2}$
Learning rate search range	$1 \times 10^{-4} - 1 \times 10^{-2}$
Batch size search range	16, 32, 64
Validation method	Repeated five-fold cross-validation
Number of folds	5
Number of repeats	3

Supplementary Table 2. Essential parameters and settings used in the NSGA-II model

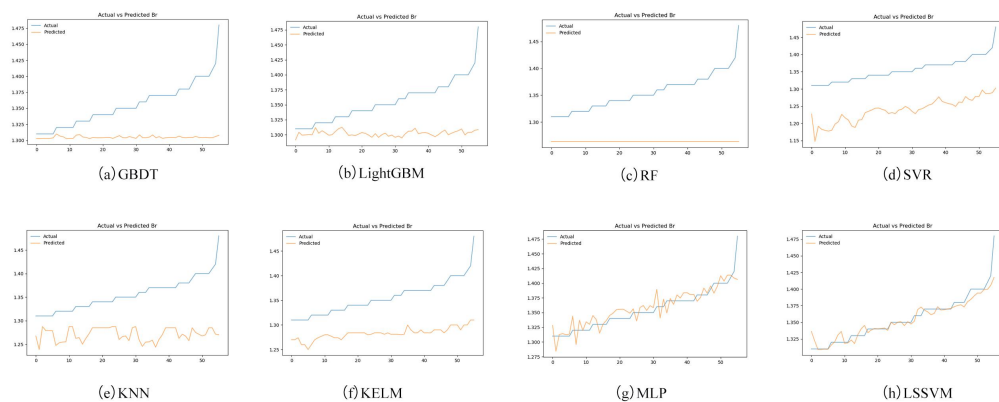
Parameter	Setting
Number of decision variables	10
Number of objectives	3
Lower bounds (varmin)	[0,0,0,0,0, 1,10,0,0,1]
Upper bounds (varmax)	[1,1,1,1,1,35,100,10,0.01,20]
Population size	npop = 700
Maximum iterations	maxit = 500
Crossover probability	pc = 0.7
Number of offspring (nc)	round(pc*npop/2)*2
Arithmetic factor	gamma = 0.1
Mutation probability	mu = 0.3
Step size	sigma = 0.05
First 5 vars perturbation	delta = 0.1*randn
Nonnegativity enforcement	max(y2,0)

Supplementary Table 3. Statistics of the datasets

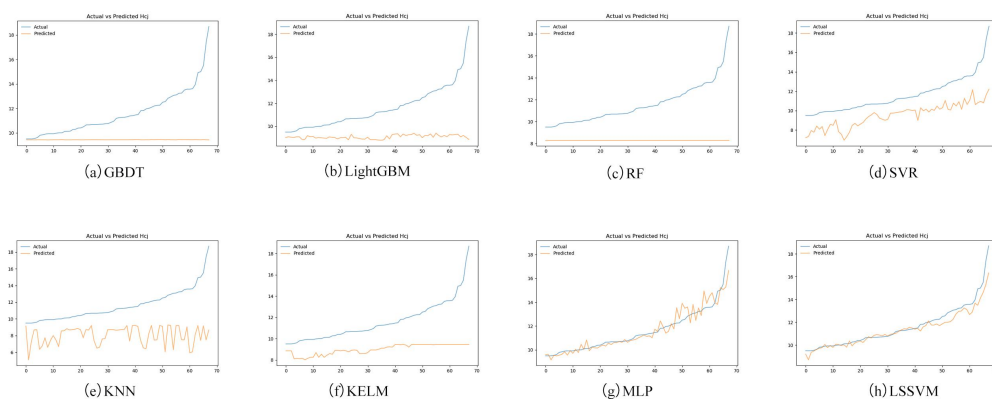
Variables		Min	Max	Mean	Median	Standard deviation
Input variables	P _{Ce}	0	1	0.12	0.08	0.15
	P _{La}	0	1	0.36	0.26	0.32
	P _Y	0	1	0.10	0	0.26
	P _{Pr}	0	1	0.04	0.05	0.11
	P _{Nd}	0	1	0.37	0.15	0.37
	REs	7	33.50	15.40	13.7	5.32
	P _{Fe}	60	91	78.17	80	5.08
	P _{Co}	0	17	0.24	0	1.75
	P _{TM}	0	8.50	0.28	0	0.89
	P _B	0.95	24	5.95	6.05	2.68
Output variables	Br	0.20	1.48	0.97	1.15	0.39
	Hcj	0.11	18.70	5.60	5.03	3.94
	(BH) _{max}	0.19	48.10	16.51	7.56	15.91

Supplementary Table 4. Performance evaluation of the MLP models for predicting Br, H_{cj}, and (BH)_{max} under different train-test splitting ratios

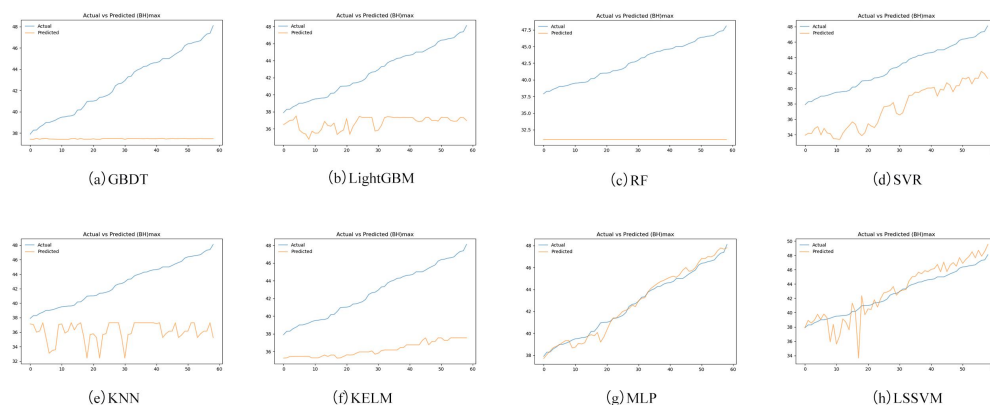
Target property	Split ratio (Train/Test)	Train R ²	Test R ²	Cross-validation R ² (Mean ± SD)
Br	70/30	0.969	0.967	0.910 ± 0.065
	80/20	0.970	0.979	0.936 ± 0.027
	90/10	0.980	0.982	0.941 ± 0.034
H _{cj}	70/30	0.950	0.866	0.841 ± 0.086
	80/20	0.950	0.901	0.826 ± 0.068
	90/10	0.942	0.934	0.821 ± 0.074
(BH) _{max}	70/30	0.695	0.668	0.816 ± 0.112
	80/20	0.978	0.955	0.850 ± 0.091
	90/10	0.960	0.971	0.857 ± 0.082



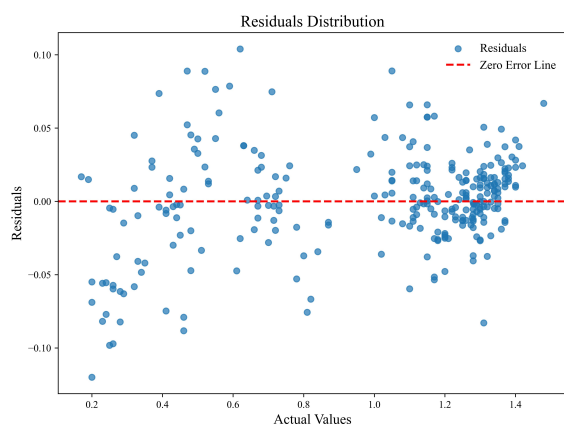
Supplementary Figure 1. Performance comparison of eight machine learning models for $(BH)_{\max}$ prediction (actual vs. predicted values).



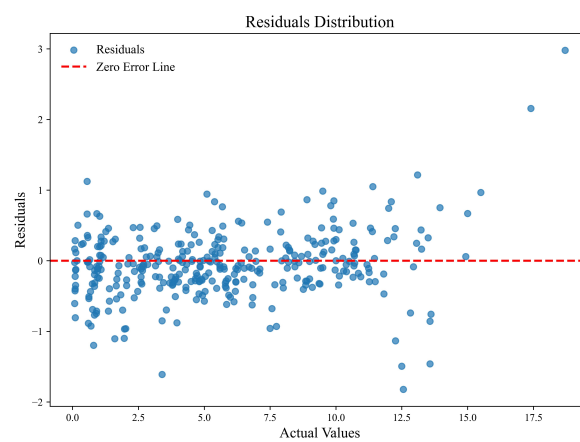
Supplementary Figure 2. Performance comparison of eight machine learning models for Br prediction (actual vs. predicted values).



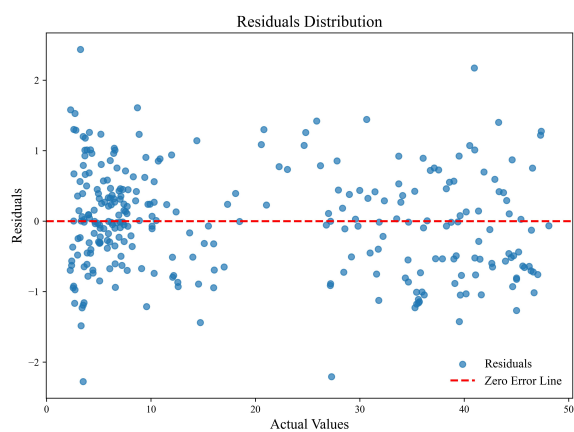
Supplementary Figure 3. Performance comparison of eight machine learning models for H_{cJ} prediction (actual vs. predicted values).



A



B



C

Supplementary Figure 4. Distribution of MLP prediction residuals for in Figure 4A-C.
(A) Br; (B) Hcj; (C) (BH)max.