

## Predicting Copper Leaching from Slag: An Interpretable Machine Learning Approach under Oxidative Sulfuric Acid Conditions

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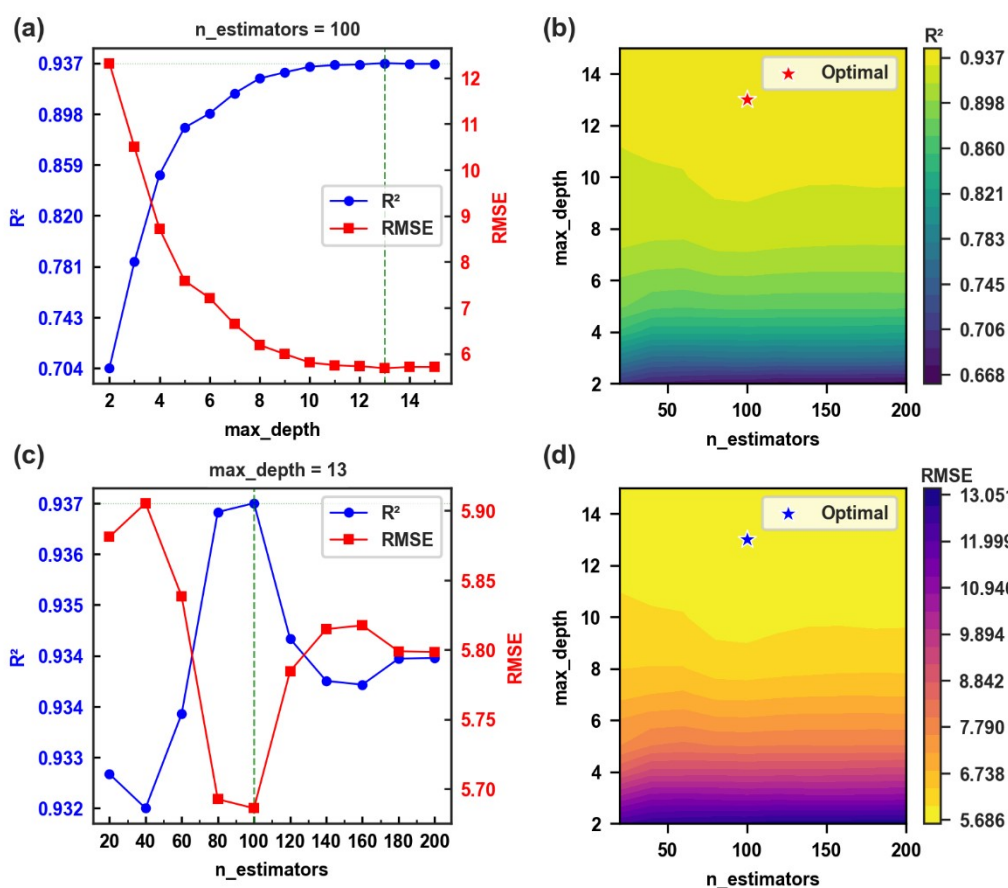


Fig.S1 (a) Optimization for the parameter of  $\max\_depth$  of the RF model; (b) Optimization for the parameter of  $n\_estimator$  of the RF model; (c) Dependence of  $R^2$  with the  $\max\_depth$  and  $n\_estimators$ ; (d) Dependence of RMSE with the  $\max\_depth$  and  $n\_estimators$

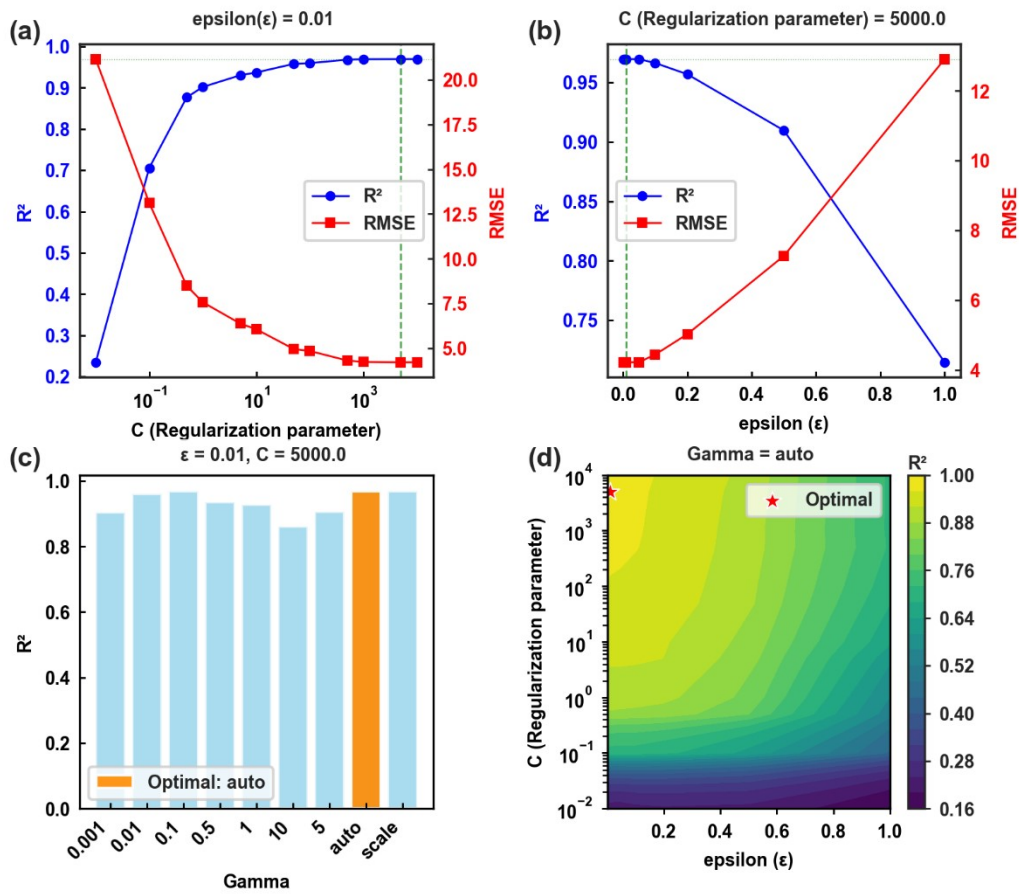


Fig.S2 (a) Optimization for the parameter of C of the SVR model; (b) Optimization for the parameter of epsilon of the SVR model; (c) Optimization for the parameter of gamma of the SVR model; (d) Dependence of  $R^2$  with the C and epsilon

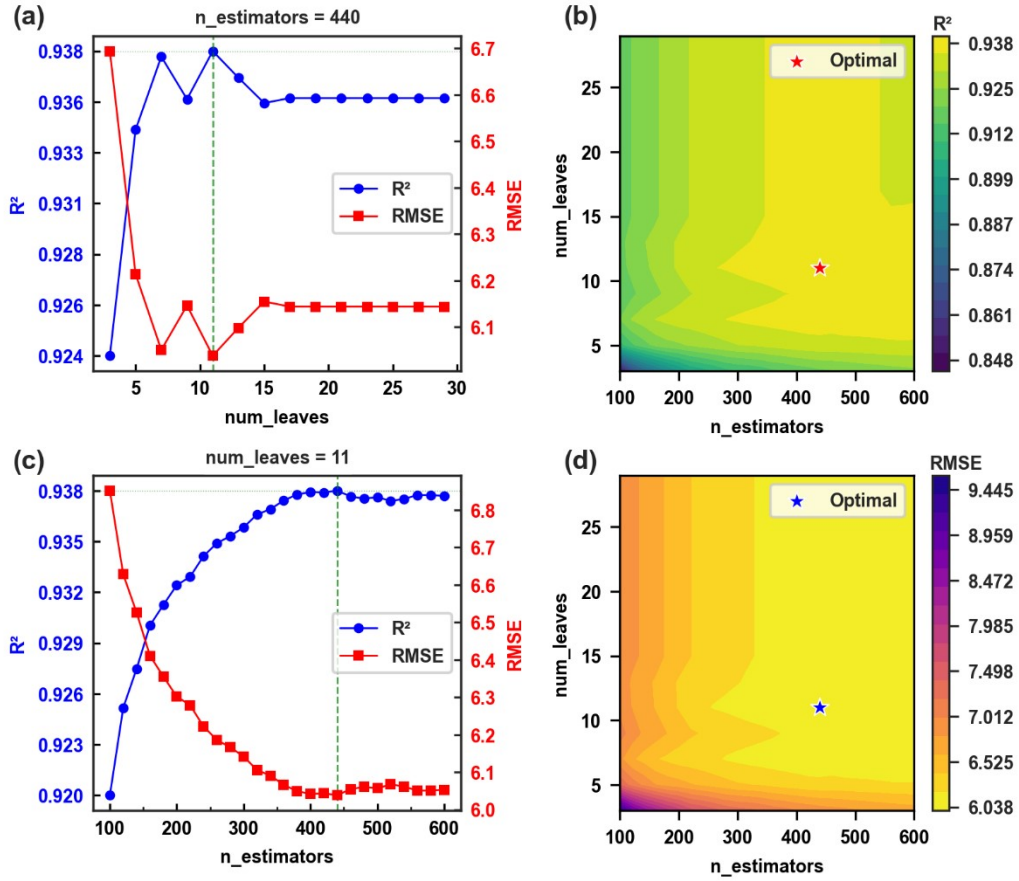


Fig.S3 (a) Optimization for the parameter of  $num\_leaves$  of the LightGBM model; (b) Optimization for the parameter of  $n\_estimator$  of the LightGBM model; (c) Dependence of  $R^2$  with the  $max\_depth$  and  $n\_estimators$ ; (d) Dependence of RMSE with the  $max\_depth$  and  $n\_estimators$

Table S1. Hyperparameter search ranges and optimal values obtained through grid-search optimization.

Model	Hyperparameter	Search range	Optimal value
XGBoost	$max\_depth$	2-15	4
	$n\_estimators$	20-200	150
Random Forest	$max\_depth$	2-15	13
	$n\_estimators$	20-200	100
LightGBM	$num\_leaves$	3-30	11
	$n\_estimators$	100-600	440
SVR	C	0.01-10000	5000
	epsilon	0.001-1.0	0.01
	gamma	auto, scale, 0.001-10	auto

Table S2. Training times of the four machine learning models under identical

conditions.

Model	Training time (s)
XGBoost	0.53
Random Forest	0.51
LightGBM	0.33
SVR	1.26

Table S3. Recommended operating ranges for copper slag leaching parameters derived from SHAP partial dependence analysis of the XGBoost model.

Parameter	Recommended range	SHAP effect / interpretation
Leaching time (min)	40-80	Positive influence; SHAP increases rapidly then plateaus beyond 80 min
Temperature (°C)	120-160	Positive; monotonic increase, plateau at high end
Acid concentration (mol/L)	0.5-1.5	Positive; sharp increase below 0.5, plateau beyond 1.5
Oxygen pressure (kPa)	100-1000	Positive; effect diminishes above 1000
Pulp density (%)	5-30	Negative trend; lower pulp density favors efficiency
Particle size (µm)	<75	Negative trend; smaller particle size favors efficiency