

Table S1. Features for the phase parametric candidates of machine learning models.

Parameter	Calculation Formula	Description	Parameter	Calculation Formula	Description
δ	$\sqrt{\sum_{i=1}^n C_i \left(1 - \frac{r_i}{r}\right)^2}$	Differences in Atomic Radii	T_m	$\sum_{i=1}^n C_i T_i$	The melting point calculated by the mixing law
χ	$\sqrt{\sum_{i=1}^n C_i (\chi_i - \bar{\chi})^2}$	Electronegativity difference	a_m	$\sum_{i=1}^n C_i a_i$	The lattice constant calculated by the mixing rule
VEC	$\sum_{i=1}^n C_i \text{VEC}_i$	Valence electron concentration	ΔT_m	$\sqrt{\sum_{i=1}^n C_i (T_i - \bar{T})^2}$	Melting point difference
ΔH_{mix}	$\sum_{i=1, i \neq j}^n 4H_{ij} C_i C_j$	Mixed Flame	a	$\sqrt{\sum_{i=1}^n C_i (a_i - \bar{a})^2}$	Melting point difference
ΔS_{mix}	$-R \sum_{i=1}^n (C_i \ln C_i)$	Mixed Xiu	ρ	$\sum_{i=1}^n C_i \rho_i$	Melting point difference
Λ	$\Delta S_{\text{mix}} / (\delta * \delta)$	Λ parameter	δ_{con}	$\sqrt{\sum_{i=1}^n C_i \left(1 - \frac{r_{\text{coni}}}{r_{\text{con}}}\right)^2}$	Covalent radius difference
γ	$\frac{\sqrt{\frac{(r+r_{\text{min}})^2 - r^2}{(r+r_{\text{min}})^2}}}{\left(1 - \frac{(r+r_{\text{max}})^2 - r^2}{(r+r_{\text{max}})^2}\right)}$	γ parameter	I	$\sum_{i=1}^n C_i I_i$	Ionization energy

D. χ	$\sum_{i=1}^n \sum_{j=1, i \neq j}^n C_i C_j$	Local	A_r	$\sum_{i=1}^n C_i A_{ri}$	Relative atomic
	$* \chi_i - \chi_j $	electronegativity mismatch			mass
E_c	$\sum_{i=1}^n C_i * (Ec)_i$	Cohesion	EA	$\sum_{i=1}^n C_i EA_i$	Electronic affinity work
λ	$\frac{\Delta S_{mix}}{\delta^2}$	Geometric parameters			

Table S2. Features for the mechanical parametric candidates of machine learning models.

Parameter	Calculation Formula	Description	Parameter	Calculation Formula	Description
The modulus					
η	$\sum_{i=1}^n \frac{C_i * \frac{2(G_i - G)}{G_i + G}}{1 + 0.5 * \left C_i * \frac{2(G_i - G)}{G_i + G} \right }$	in the	δG	$\sqrt{\sum_{i=1}^n C_i * \left(1 - \frac{G_i}{G}\right)^2}$	Difference in
		reinforcement model does not match			shear modulus
D. r	$\sum_{i=1}^n \sum_{j=1, i \neq j}^n C_i C_j * r_i - r_j $	Partial size mismatch	D. G	$\sum_{i=1}^n \sum_{j=1, i \neq j}^n C_i C_j$	Local modulus
				$* G_i - G_j $	mismatch
A	$G * \delta r * (1 + \mu)(1 - \mu)$	Strengthen the	G	$\sum_{i=1}^n C_i * G_i$	Shear
		energy term in the model			modulus

F	$\frac{2G}{1 - \mu}$	Peierls- Nabarro factor	E	$\sum_{i=1}^n C_i * E_i$	Young's modulus
μ	$\frac{1}{2} E * \delta$	Lattice distortion energy	ν	$\frac{E_i}{2G_i} - 1$	Poisson's ratio

Table S3. The specifically designed parameters of Gaussian Processes models.

Output	Parameters
System Energy	'Crossval', 'on'; 'Fitmethod', 'exact'; 'ComputationMethod', 'qr'; 'Sigma', 0.025; 'Regularization', 0.1; 'PredictMethod', 'exact'
Cell Volume	'Crossval', 'on'; 'Fitmethod', 'exact'; 'ComputationMethod', 'qr'; 'Sigma', 0.025; 'Regularization', 0.01; 'PredictMethod', 'exact'
Young's Modulus	'Crossval', 'on'; 'Holdout', 0.3; 'Fitmethod', 'exact'; 'ComputationMethod', 'qr'; 'Standardize', true; 'BasisFunction', 'linear'; 'KernelFunction', 'ardsquaredexponential'; 'PredictMethod', 'exact'; 'ConstantSigma', true; 'Sigma', 0.02; 'Regularization', 0.06; 'BasisFunction', 'constant'
Bulk Modulus	'Crossval', 'on'; 'Holdout', 0.2; 'Fitmethod', 'sr'; 'ComputationMethod', 'qr'; 'Standardize', true; 'KernelFunction', 'ardsquaredexponential'; 'PredictMethod', 'sr'; 'ConstantSigma', true; 'Sigma', 0.003; 'BasisFunction', 'linear'
Shear Modulus	'Crossval', 'on'; 'Fitmethod', 'none'; 'KernelFunction', 'exponential'; 'BasisFunction', 'none'; 'ComputationMethod', 'v'; 'ConstantSigma', false; 'Regularization', 0.01; 'PredictMethod', 'sr'

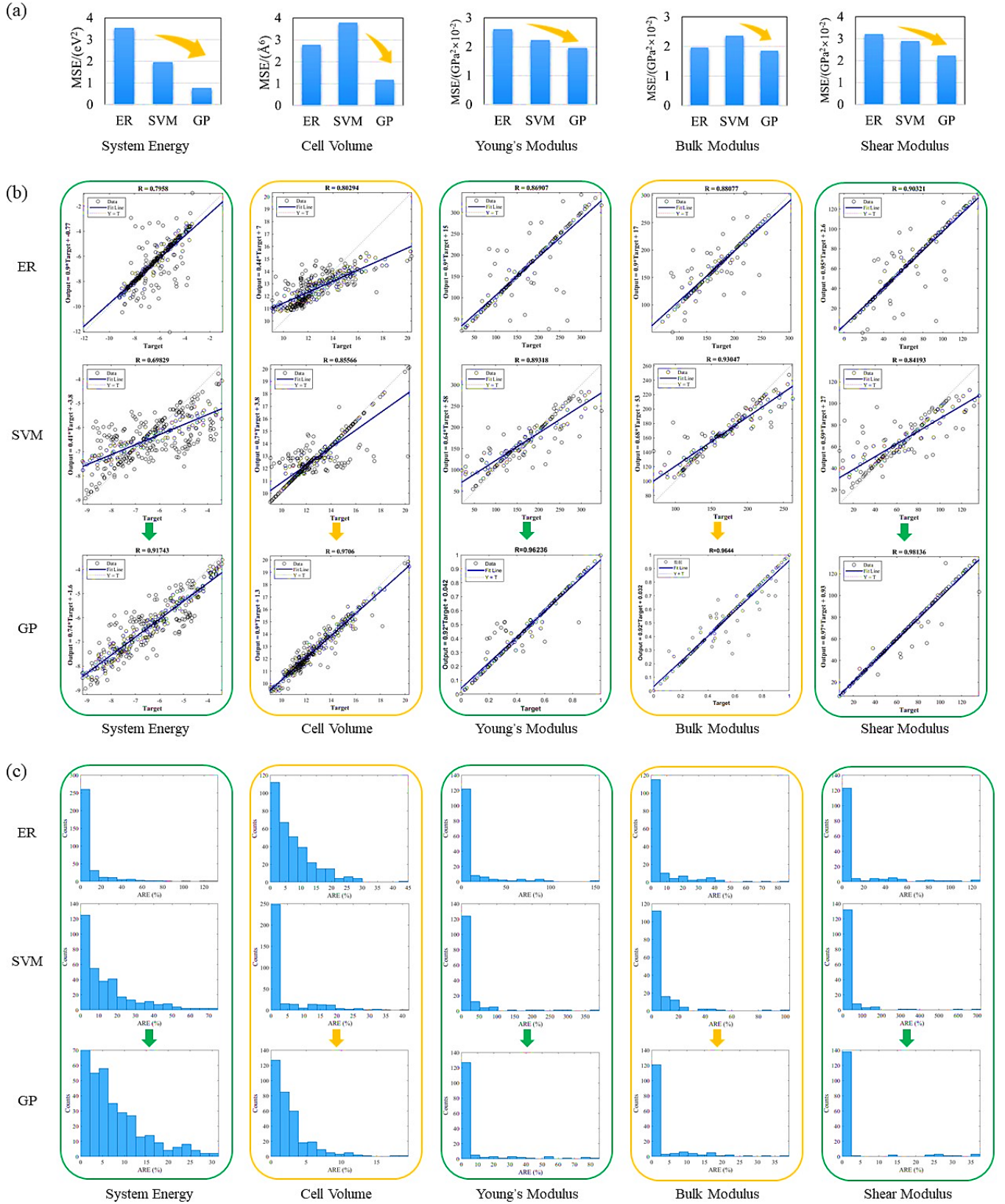


Figure S1. The (a) MSE, (b) regression figures and (c) ARE of five outputs ML models trained with ER, SVM and GP algorithms.