

Machine learning enables the discovery of 2D *Invar* and anti-*Invar* monolayers

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Materials demonstrating positive thermal expansion (PTE) or negative thermal expansion (NTE) are quite common, whereas those exhibiting zero thermal expansion (ZTE) are notably scarce. In this work, we identify the mechanical descriptors, namely in-plane tensile stiffness and out-of-plane bending stiffness, that can effectively classify PTE and NTE 2D crystals. By utilizing high throughput calculations and the state-of-the-art symbolic regression method, these descriptors aid in the discovery of ZTE or 2D *Invar* monolayers with the linear thermal expansion coefficient (LTEC) within $\pm 2 \times 10^{-6} \text{ K}^{-1}$ in the middle range of temperatures. Additionally, the descriptors assist the discovery of large PTE and NTE 2D monolayers with the LTEC larger than $\pm 15 \times 10^{-6} \text{ K}^{-1}$, which are so-called 2D anti-*Invar* monolayers. Advancing our understanding of materials with exceptionally low or high thermal expansion is of substantial scientific and technological interest, particularly in the development of next-generation electronics at the nanometer or even Ångstrom scale.

The dimensions of the solids undergo reversible expansion or dilatation at a macroscopic level when they are heated or cooled due to alterations in the thermal vibrations inherent within their crystal structure, which can be measured by the linear coefficient of thermal expansion (α , LCTE)^{1,2}. Most materials exhibit positive thermal expansion (PTE) coefficients, which leads to expanded lattices with increasing temperature due to the population of high-energy anharmonic phonons³. While uncommon, some materials expand when cooled, demonstrating negative thermal expansion (NTE). Even more exceptional is zero thermal expansion (ZTE, $\alpha \approx 0$), also called the *Invar* effect, where the material neither expands nor contracts volumetrically within specific temperature ranges^{2–8}. Since the first discovery of the *Invar* in 1897 by Guillaume⁹ and then won the Nobel Prize in Physics in 1920, some *Invar* materials have been found, including ferromagnetic alloys, ferroelectric relaxors, silica-based glasses and crystals with nano-porous structure^{2,6–8,10–12}. The *Invar* materials are widely used in temperature-sensitive devices and high antithermal shock applications, such as the construction of calibrated, high-precision mechanical and optical instruments, such as seismographs, telescope technology on Earth and in space, spring watches with atomic time standards, heat-engine components and so on^{2,5}. Materials

with very low thermal expansion are not only crucial for practical applications but also fascinating scientifically.

In recent decades, substantial advancements in the field of nanofabrication have contributed to the reduction in the size of electronic devices to unprecedented dimensions, reaching scales as small as nanometers and even extending to Ångstrom levels^{13–16}. The 2D material monolayers can be used as ideal building blocks to create devices and instruments with the size of nanometers or even Ångstrom. The thermal expansion properties of 2D materials show significant importance in electronic and optical applications since the accumulated thermal stress/strain would degrade the performance and stability under varying temperatures¹⁷. The reported α for 2D crystal either shows PTE (like MoS₂) or NTE (like graphene)^{18,19}. However, by now, no 2D crystal monolayer with ZTE or 2D *Invar* has been discovered. Recently developed 2D materials databases, such as C2DB, 2DMatPedia, and JARVIS, contain more than 10000 2D materials^{20–22}. It makes us use data mining and machine learning methods to discover and understand the materials with remarkable properties, rather than study materials case by case^{23–26}. Considering the existence of both PTE and NTE 2D crystals, we believe the presence of ZTE counterparts among these materials.

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Recently, employing active learning, two high-entropy alloys (*bulk Invar*) featuring remarkably low α of $-2 \times 10^{-6} \text{ K}^{-1}$ were identified¹⁰. In this work, the art-of-state machine learning methods of classification and symbolic regression are used to advance the understanding and discovery of *2D Invar* monolayers. A general magnetism for Invar materials is that a negative thermal expansion coefficient counteracts a PTE of the lattice. Herein, we identify the mechanical descriptors for thermal expansion from extensive data, that is in-plane tensile stiffness and out-of-plane bending stiffness. These two descriptors demonstrate effectiveness in the classification of PTE and NTE 2D crystals. By utilizing high throughput calculations and the state-of-the-art symbolic regression method, these descriptors aid in the discovery of 2D crystals with exceptionally zero, positive, and negative thermal expansion. We predict the existence of ZTE 2D crystals or 2D Invar materials with remarkably low LTEC in the middle range of temperatures, following the criteria $|\alpha| < 2 \times 10^{-6} \text{ K}^{-1}$ proposed in the review paper of R. Roy and co-workers². We also discovered 2D crystals with extremely large thermal expansion (ELTE), that is extreme PTE (EPTE) and NTE (ENTE), which is called the *anti-Invar effect*²⁷, with α larger than $\pm 15 \times 10^{-6} \text{ K}^{-1}$. Enhancing our comprehension of materials exhibiting exceptionally low or high thermal expansion holds significant scientific and technological promise, notably in broadening the scope of applications for 2D materials in the post-Moore era¹³.

Results

In-plane and out-of-plane deformation modes of 2D crystals

At a finite temperature, a free-standing two-dimensional material sheet exhibits both in-plane and out-of-plane deformation modes²⁸. For simplification, we employ a one-dimensional (1D) chain model to illustrate this question, as depicted in Fig. 1a. The in-plane mode is determined by the intrinsic expansion of chemical bonds that the length is enlarged with the rising of temperature²⁹, thus resulting in a positive contribution to the thermal expansion. Conversely, the out-of-plane mode due to the thermal fluctuations contributes to the in-plane contraction, which has been identified as the origin of NTE behavior³⁰. Figure 1b illustrates the atomic configurations of graphene, showcasing the ripples induced by thermal fluctuations. Under the competition between the in-plane and out-of-plane modes, 2D crystal monolayers exhibit diverse thermal expansion behaviors. For example, the graphene exhibits NTE behavior that $\alpha = -3.33 \times 10^{-6} \text{ K}^{-1}$ at 300 K and $\alpha < 0$ up to 2500 K (as shown in Fig. 1c)¹⁹. While MoS₂ is PTE from about 20–1000 K that α increases with temperature and $\alpha = 7.26 \times 10^{-6} \text{ K}^{-1}$ at 600 K.

Data-driven classification of PTE and NTE 2D crystals

To achieve rational classifications of PTE and NTE 2D crystals, we take into account pre-existing physical knowledge to select physically meaningful features, including thickness (h), volume per atom (vpa),

average bond length (abl), average atomic radius (aar), bond density (bd), in-plane stiffness E_{2D} and out-of-plane bending stiffness D . By using random forest method³¹, we find the most important features for thermal expansion coefficient of 2D crystal monolayers are E_{2D} and D as shown in Supplementary Fig. 1. The result is reasonable that these two mechanical properties describe the ability of 2D crystals to resist in-plane and out-of-plane deformation as mentioned above, respectively. To confirm the rationality of E_{2D} and D to the thermal expansion of 2D crystals, we employ the recently developed sure independence screening sparsifying operator (SISSO) method, which is a compressed sensing-based methodology that can generate predictive models and classification descriptors³². We trained a SISSO classification model to separate PTE and NTE 2D crystals using E_{2D} and D (Fig. 2a). Two descriptors (S) are obtained, that is $S_1 = 1/D$ and $S_2 = 1/E_{2D}^2$. Herein, the separation line is obtained by a linear support vector machine (SVM) and all of the data are perfectly classified by the classification model. We find that the materials with similar S_1 (or S_2), the farther away from the classification line, the larger of the value of α is. For example, the E_{2D} of WSe₂ (115.81 N/m) is similar to SnO₂ (119.34 N/m), but the D of WSe₂ (11.05 eV) is 2.73 times of SnO₂ (4.05 eV). Thus WSe₂ is farther from the separation line than SnO₂ (see Fig. 2a), and we find the α of WSe₂ ($8.43 \times 10^{-6} \text{ K}^{-1}$ at 500 K) is larger than SnO₂ ($4.07 \times 10^{-6} \text{ K}^{-1}$ at 500 K). Figure 2b shows the temperature dependence of the LTECs for 2H phase MoS₂, MoSe₂, and MoTe₂ and their positive LTECs follow the order of MoS₂ < MoSe₂ < MoTe₂. It can be understood that the E_{2D} follows the order of MoS₂ (123.12 N/m) > MoSe₂ (103.26 N/m) > MoTe₂ (78.50 N/m), while the bending stiffness D follows the order of MoS₂ (8.27 eV) < MoSe₂ (9.59 eV) < MoTe₂ (10.37 eV) (see inset in Fig. 2b). Therefore, the classification model can assist us in understanding thermal expansion of 2D crystals and discovering that with either significantly large or zero thermal expansion. The EPTE 2D crystals should be distributed far below the separation line, demonstrating low E_{2D} and high D . Similarly, the ENTE 2D crystals should be situated significantly above the separation line, exhibiting high E_{2D} and low D . The ZTE materials are expected to be distributed near the separation line as the balance of in-plane and out-of-plane deformation.

The interpretation of mechanical descriptors

The descriptors obtained by the SISSO method offer distinct physical insights and are explainable. The descriptor S_1 represents the in-plane constriction caused by out-of-plane thermal fluctuations (as shown in Fig. 1a). Considering a square sheet with the size $L_0 \times L_0$, the in-plane contraction can be quantified by the bending energy based on the statistical mechanics of membranes (see detailed derivation in SI)³³. The bending energy of bend configuration can be expressed in terms of the Fourier coefficients as^{34,35}, $U_b = \frac{L_0^2}{2} \sum D|\mathbf{q}|^4 |\bar{w}(\mathbf{q})|^2$, where \mathbf{q} is wave-number vector in 2D space, $\bar{w}(\mathbf{q})$ is the Fourier coefficients. The probability density function of each possible bending configuration

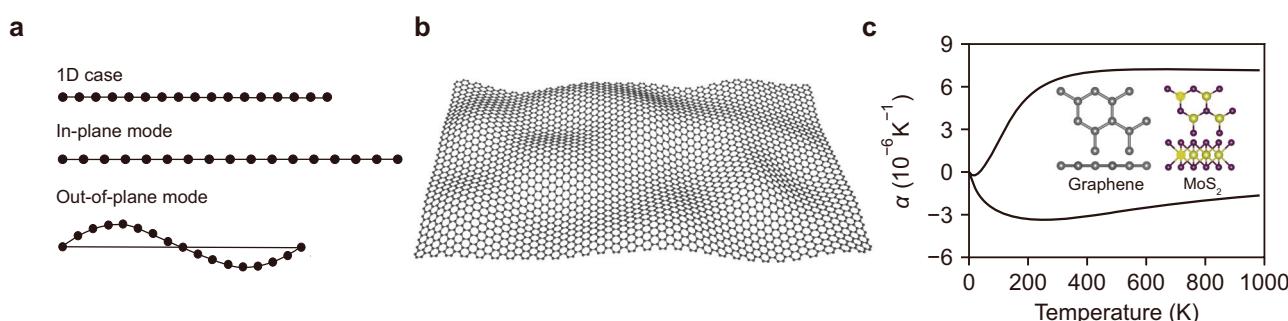


Fig. 1 | In-plane and out-of-plane deformation modes of 2D crystals. **a** A one-dimensional atomic chain model for the in-plane and out-of-plane deformation. **b** The atomic configurations of the ripple of graphene membrane. **c** The linear

thermal expansion coefficient (α) of graphene and MoS₂(2H) at temperatures ranging from 0 K to 1000 K. The inset shows the crystal structures of graphene and MoS₂.

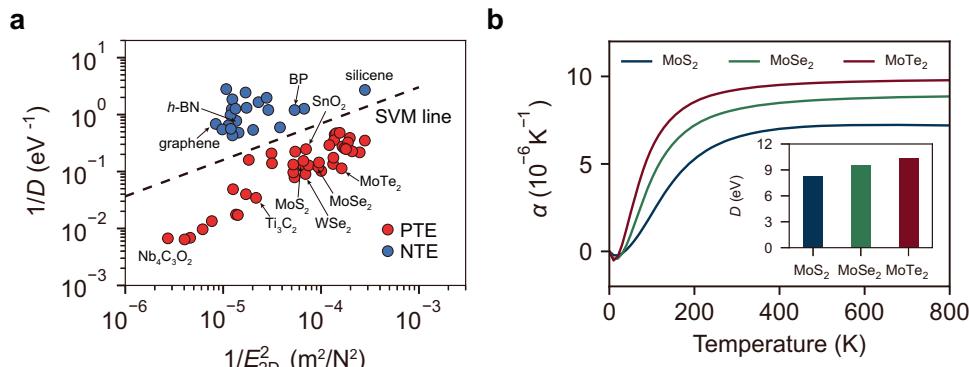


Fig. 2 | Data-driven classification of PTE and NTE 2D crystals. **a** The positive thermal expansion (PTE)/negative thermal expansion (NTE) classification map for 2D crystals (in logarithmic coordinate) using in-plane stiffness E_{2D} and bending stiffness D . All of the training data are well classified into two groups (blue dots represent NTE materials and red dots represent PTE materials at the temperature range of 300–1000 K). The separation line is obtained by linear support vector

machine (SVM) with descriptors found by sure independence screening sparsifying operator (SISSO). **b** The LTECs (α) of MoS_2 , MoSe_2 , MoTe_2 (here are all 2H phases) as a function of temperature. The inset shows their D . All the data of E_{2D} and D can be found in Supplementary File. Unless specified otherwise, all transition metal dichalcogenides (TMDs) discussed in this work are the 2H phase.

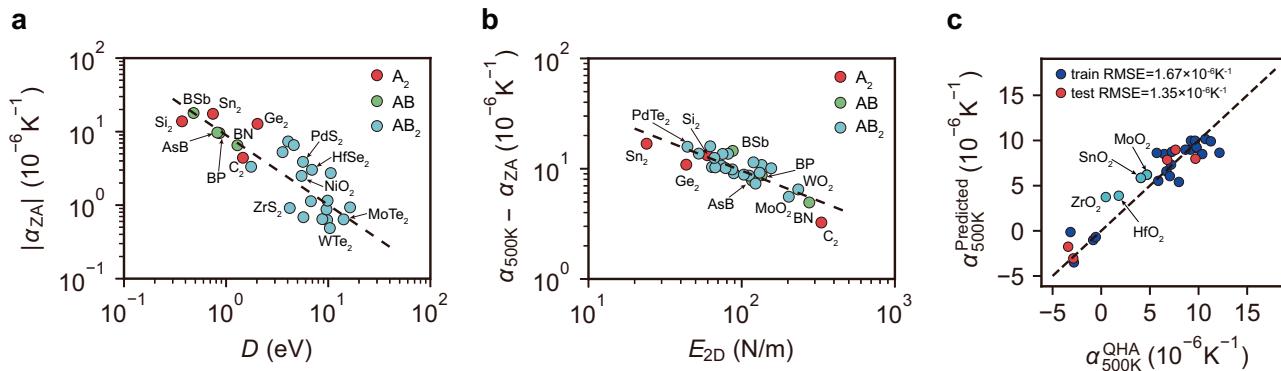


Fig. 3 | Phonon mode contribution and LTEC prediction. **a** The relation between the contribution of ZA mode ($|\alpha_{ZA}|$) and bending stiffness (D). **b** The relation between $\alpha_{500K} - \alpha_{ZA}$ and in-plane stiffness E_{2D} . Here α_{500K} is LTECs at 500 K. The black dash lines in (a) and (b) are obtained by fitting. **c** The predicted α_{500K} against QHA results. The black dash line is diagonal. The values of E_{2D} , D , and α_{500K} in (a)

and (b) are summarized in Supplementary Table 1 and Supplementary File. Here 80% of the data are randomly selected for training and 20% for validation. Herein, we also add the data of zero thermal expansion (ZTE) 2D monolayers and the data that the LTEC is relatively small but just out of the range of ZTE on the figure.

can be given by Boltzmann distribution as $\rho_i \propto \exp(-U_b/k_B T)^{36}$, and the expectation of out-plane contraction strain is $\langle \varepsilon_x \rangle = \mu_n k_B T / 16\pi^2 D$, where μ_n is a dimensionless parameter, k_B is the Boltzmann constant, and T is the temperature. Therefore, $\alpha \sim \langle \varepsilon \rangle \sim 1/D$. The descriptor S_2 represents in-plane thermal expansion as a result of anharmonic effect³³. Assuming the out-of-plane modes are completely suppressed (that is pure 2D deformation), the in-plane thermal expansion coefficient (α_{2D}) of a solid depends on the intrinsic thermal expansion coefficient of internal bonds α_b ^{37,38}. According to the empirical relationship proposed by H. Megaw³⁹, the thermal expansion coefficients of bonds α_b correlate to the valance of a bond (s), that is $\alpha_b \propto s^{-2}$. In addition, the force constant of a bond k_b is related to bond valance $k_b \propto s$ (see more discussion in SI)³⁸. The in-plane thermal strain can be expressed as $\varepsilon_{2D} = \alpha_{2D} T = 3.02 \times \beta T / (E_{2D})^2$ (the unit of β is $\text{K} \times \text{m}^2/\text{N}^2$). Therefore, $\varepsilon_{2D} \sim 1/E_{2D}^2$.

For 2D crystals, the flexural phonon mode (ZA) will be abnormally hardened under tensile strain (see Supplementary Fig. 2), leading to the negative mode Grüneisen parameter γ , and the negative γ of ZA mode is the main contributor to NTE⁴⁰. The contribution of ZO mode is also negative but much smaller than that of ZA mode, and all the other phonon modes associated with in-plane motion have positive γ ⁴¹. To examine the branch contribution of ZA modes, we employ the Grüneisen theory (GT)¹, as the branch contribution cannot be obtained

simply through the QHA while easily by GT. This approach demonstrates validity comparable to QHA (details in Supplementary Fig. 3) but with significantly enhanced computational efficiency. Herein, we analysed the contribution of ZA modes for 28 2D crystals to the LTEC at 500 K (α_{500K}). Figure 3a shows that the contribution of ZA mode ($|\alpha_{ZA}|$) is negatively correlated with D , and we fitted the relation between $|\alpha_{ZA}|$ and D by $|\alpha_{ZA}| = \zeta_1 \times D^{\beta_1}$ that the resulting $\beta_1 = -0.924$ and $\zeta_1 = 9.2 \times 10^{-6} \text{ K}^{-1}/\text{eV}^{\beta_1}$, which is similar with the descriptor S_1 obtained by SISSO method. Therefore, S_1 could represent the impact of thermal fluctuations on the thermal expansion. Other than the ZA branch, the contribution of the remaining phonon modes to the thermal expansion coefficient is represented as $\alpha - \alpha_{ZA}$. Here, we take the result of 500K as an example. As seen clearly from Fig. 3b that $\alpha_{500K} - \alpha_{ZA}$ negatively correlates with E_{2D} , and the relation between $\alpha_{500K} - \alpha_{ZA}$ and E_{2D} is fitted that $\alpha_{500K} - \alpha_{ZA} = \zeta_2 \times E_{2D}^{\beta_2}$ ($\beta_2 = -0.545$, $\zeta_2 = 117.49 \times 10^{-6} \text{ K}^{-1}/(\text{N}/\text{m})^{\beta_2}$), which exhibits the same trend as the descriptor S_2 , indicating that S_2 describes the in-plane expansion of 2D materials due to the expansion of internal chemical bonds.

The discovery of ZTE 2D crystals

Although the data of E can be calculated from elasticity tensors (C_{ij}) that are included in C2DB, the data of D is absent and the calculation of the D is expansive. Therefore, we first predict the D by symbolic

regression (see details in Supplementary Fig. 4). Six features associated with D are selected, including E_{2D} , density (ρ), volume per atom (vpa), thickness (h), average bond length (abl) and bond density (bd). We trained a regression model to predict D based on these descriptors and presented the results in Supplementary Fig. 4 with the root mean squared error (RMSE) of 1.34 eV. The results would be improved if more training data was used. According to the theory of thermal expansion for bulk solids^{42,43}, the thermal expansion is related to the volume of the unit cell (V), the isothermal compressibility (K_T), that is $\alpha = K_T/V$. Therefore, we choose quantities along with E_{2D} and D as input features to predict the LTEC at 500K (α_{500K}) and the expression generated by symbolic regression is $\alpha_{500K} = (14.31 \times K_T - \frac{0.038}{D}) \times E_{2D} + 14.27 \times \frac{D}{V}$. Here all the quantities are just the *number* and the units can be found in Supplementary Table 1. The predicted α_{500K} is plotted against the QHA results as shown in Fig. 3c. The RMSE is $1.67 \times 10^{-6} \text{ K}^{-1}$ ($1.35 \times 10^{-6} \text{ K}^{-1}$) and the coefficient of determination (R^2) is 0.91 (0.93) for the training (testing) set, which is accurate enough for the rough screening of ZTE. Herein, we also add the data of ZTE 2D monolayers and the data that the LTEC is relatively small but just out of the range of ZTE on the figure. The results show these data are also near the diagonal, which confirms the regression model is robust. From the expression of α_{500K} , we observed that α_{500K} increases with the increase of D , which is consistent with our discussion above.

According to the conclusions from the classification model, ZTE materials should possess comparable in-plane and out-of-plane modes. By using the regression model of α_{500K} with reliable predictive results, we can screen the ZTE monolayers (see details in Fig. 4). By using the criteria of stability proposed in our previous works^{23,44}, 1224 stable 2D crystals are selected from the C2DB²⁰ as candidates. Then we screen 10 potential ZTE materials with $|\alpha_{500K}| < 2\alpha_c^{ZTE}$, where $\alpha_c^{ZTE} = 2 \times 10^{-6} \text{ K}^{-1}$. It should be noted that our criterion for rough screening is moderate considering the error of prediction model. By further Grüneisen theory and QHA verification, we find 2 ZTE 2D crystals that the α can within $\pm \alpha_c^{ZTE}$ in the middle range of temperatures (300–600 K). As shown in Fig. 5, the LTECs of ZrO_2 and HfO_2 are $0.47 \times 10^{-6} \text{ K}^{-1}$ and $1.80 \times 10^{-6} \text{ K}^{-1}$ at 500 K, respectively. The minimum value of the LTECs are $-6.90 \times 10^{-6} \text{ K}^{-1}$ for ZrO_2 and $-3.16 \times 10^{-6} \text{ K}^{-1}$ for HfO_2 , and the LTECs are negative at low temperatures until they become positive at about 400 K (ZrO_2) and 170 K (HfO_2). Ultimately, they maintain ZTE behaviors over a wide temperature range.

The crystal structures of ZrO_2 and HfO_2 are shown in Fig. 5a, b, and they are categorized as 2D transition metal oxides (TMOs), characterized by their notably dense atomic configurations and reduced thicknesses (4.96 Å for ZrO_2 and 4.98 Å for HfO_2). Therefore, these TMOs typically exhibit relatively large E_{2D} (151.93 N/m for ZrO_2 and 157.27 N/m for HfO_2) and moderate D (3.67 eV for ZrO_2 and 4.48 eV for HfO_2). Apart from ZrO_2 and HfO_2 , the LTECs of TMOs are also found to be relatively small. For example, SnO_2 and MoO_2 demonstrate LTECs of $4.07 \times 10^{-6} \text{ K}^{-1}$ and $4.69 \times 10^{-6} \text{ K}^{-1}$ at 500 K (as shown in Supplementary Fig. 5), respectively. The ZTE behavior can also be understood by the Grüneisen parameters of ZA mode near the Γ point calculated as approximately -54 for ZrO_2 and -81 for HfO_2 (as shown in Supplementary Fig. 6), which is much smaller than that of well-known MoS_2 (≈ -10). The relatively smaller Grüneisen parameters lead to small LTEC values compared to MoS_2 .

The discovery of ELTE 2D crystals

Because the regression model of α_{500K} cannot predict the data located outside of the training set, we cannot use it to predict α_{500K} for ELTE. Here we refer to the classification model mentioned above (Fig. 2a). As for EPTE, it should possess large D and small E_{2D} . In other words, 2D crystals composed of multiple atomic layers and weak bonds are more likely to exhibit EPTE, because D is positively correlated with h and E_{2D} is positively correlated with k_b ^{45,46}. Guided by this rule, we screened 50 stable 2D crystals with $D > 8 \text{ eV}$ and $E_{2D} < 80 \text{ N/m}$ and calculated their LTECs using Grüneisen theory (see Fig. 4 for the detailed process). Then we screen the potential EPTE 2D crystals by $|\alpha_{500K}| > \alpha_c^{ELTE}$, where $\alpha_c^{ELTE} = 15 \times 10^{-6} \text{ K}^{-1}$, which is beyond the criteria of the high expansion groups proposed by R. Roy and co-workers ($8 \times 10^{-6} \text{ K}^{-1}$)². By further QHA verification, finally, 3 EPTE 2D crystals are found in this work and the LTECs are shown in Fig. 5j. All these materials exhibit higher LTECs than $15 \times 10^{-6} \text{ K}^{-1}$ above 500 K. Specifically, SbSe_2 possesses the highest α_{500K} ($24.49 \times 10^{-6} \text{ K}^{-1}$) among these 3 EPTE materials, which is 3.27 times of the well-known MoS_2 ($7.48 \times 10^{-6} \text{ K}^{-1}$), while HfBi_2 and HfSb_2 are 2.27 and 2.65 times that of MoS_2 , respectively. The commonality among these EPTE materials lies in their relatively large thickness (as shown in Fig. 5c–e and Table 1), resulting in their large bending stiffness D (9.03 eV for SbSe_2 , 13.74 eV for HfBi_2 and 10.01 eV for HfSb_2). The E_{2D} of EPTE materials is significantly lower than the well-known MoS_2 (123.12 N/m) (see Table 1), and the E_{2D} of SbSe_2 (31.88 N/m) is the lowest among the three, which is approximately only one-quarter of

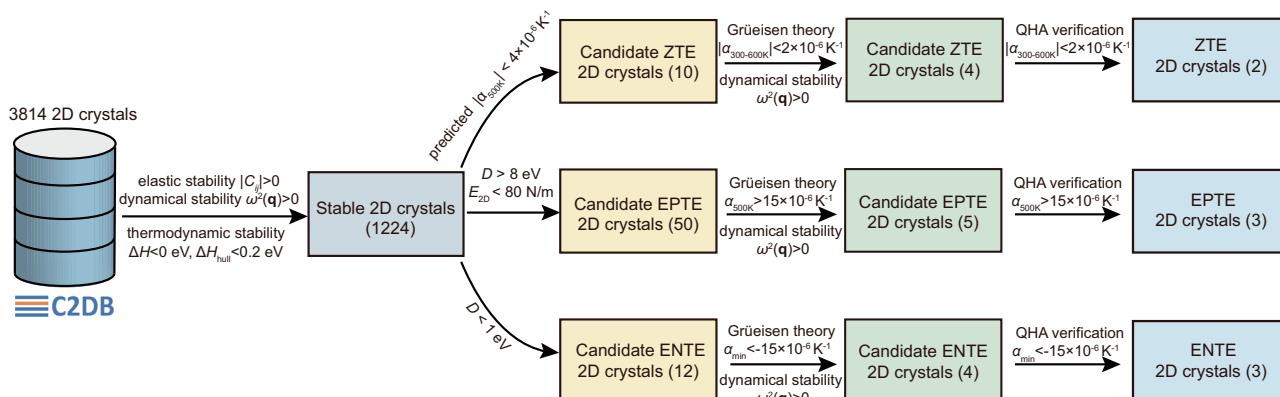


Fig. 4 | The high-throughput screening process. We selected 1224 stable 2D crystals from the C2DB database according to their elastic stability, dynamical stability (only considering the phonon frequency at high symmetric points), and thermodynamic stability. Then, we identified candidate ZTE, extreme NTE (ENTE), and extreme PTE (EPTE) 2D crystals using certain criteria ($|\alpha_{500K}| < 4 \times 10^{-6} \text{ K}^{-1}$ for ZTE using the regression model, $D > 8 \text{ eV}$ and $E_{2D} < 80 \text{ N/m}$ for EPTE, $D < 1 \text{ eV}$ for ENTE). It should be noted that our criterion for the preliminary screening of ZTE 2D

crystals is set at a moderate level ($|\alpha_{500K}| < 2\alpha_c^{ZTE}$), considering the error of the predictive model. Following this, we performed fine screening using the Grüneisen theory and dynamical stability checking (considering the phonon frequency at full q space). Finally, the results are validated by the QHA calculations. For the fine screening of ENTE, the criteria is $\alpha_{\min} < -15 \times 10^{-6} \text{ K}^{-1}$, where α_{\min} is the minimum value of the LTECs at low temperature.

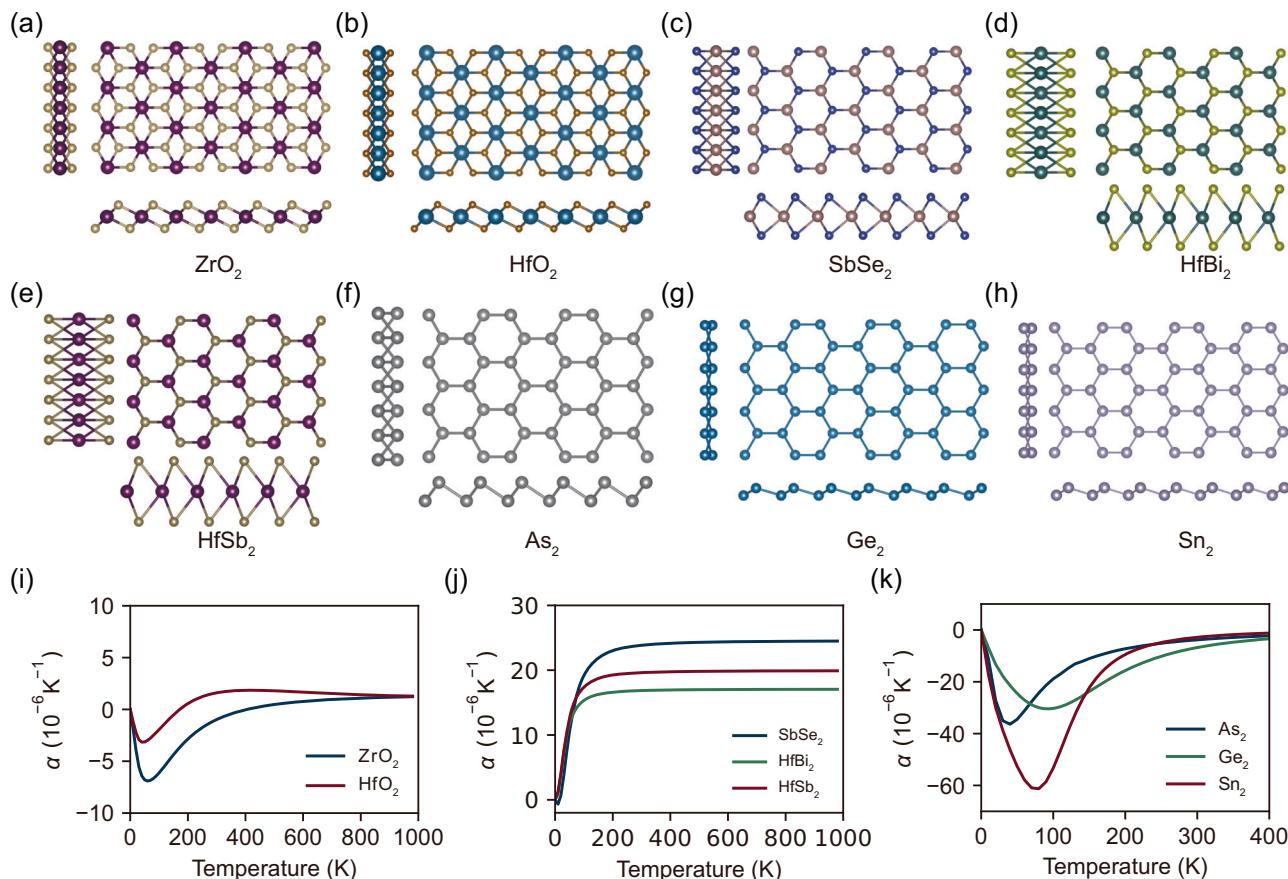


Fig. 5 | The predicted 2D Invar and anti-Invar monolayers. Crystal structures of (a) ZrO₂, (b) HfO₂, (c) SbSe₂, (d) HfBi₂, (e) HfSb₂, (f) As₂, (g) Ge₂, and (h) Sn₂. The LTECs as a function of temperature for (i) ZTE 2D materials, (j) EPTE 2D materials, and (k) ENTE 2D materials.

MoS₂, resulting in the highest LTEC of SbSe₂ among the three materials.

Correspondingly, a small value of D combined with large E_{2D} can lead to ENTE, indicating that 2D materials with a single atomic layer and strong covalent bonds are more likely to exhibit ENTE. Two famous 2D materials, that are *h*-BN and graphene with plane and hexagonal lattice, possess high E_{2D} ($>200 \text{ N/m}$) and low D ($<2 \text{ eV}$). We found that both graphene and *h*-BN demonstrate NTE behaviors below 1000 K (see LTECs in Supplementary Fig. 7). However, the minimum values of LTEC are $-3.35 \times 10^{-6} \text{ K}^{-1}$ for graphene and $-6.05 \times 10^{-6} \text{ K}^{-1}$ for *h*-BN, respectively, suggesting that they are not negative enough to be ENTE 2D crystals. Based on the discussion above, the ZA mode is the main contributor to NTE and $|\alpha_{ZA}|$ negatively correlates with D . Herein,

we lose the screening criteria to $D < 1 \text{ eV}$ considering the rarity of 2D materials with a large E_{2D} and small D . We screen 12 2D materials and 3 ENTE materials are found in this work (see Fig. 4). As seen from Fig. 5k, the LTECs of As₂, Ge₂ and Sn₂ are far below $-15 \times 10^{-6} \text{ K}^{-1}$ at low temperatures, and the minimum value of the LTECs (α_{min}) are $-36.39 \times 10^{-6} \text{ K}^{-1}$ for As₂, $-30.46 \times 10^{-6} \text{ K}^{-1}$ for Ge₂ and $-61.27 \times 10^{-6} \text{ K}^{-1}$ for Sn₂. One point should be noted that these materials only exhibit ENTE at low temperatures. This occurs because the low-lying ZA modes with negative Grüneisen parameters will be activated firstly as the temperature increases, followed by the gradual activation of other high-frequency phonon modes, which can be shown by the temperature-dependent macroscopic Grüneisen parameters in Supplementary Fig. 8. It's worth noting that some NTE 2D crystals, like Si₂ and BP (see Supplementary Table 1 and Supplementary Fig. 9), exhibit ZTE at high temperature. Our study specifically delves into the ZTE in the middle range of temperatures (300 to 600 K). Although the results of α_{500K} for these two are within in ZTE range, the α at low temperature are notably negative. Consequently, we still categorize these two materials as NTE 2D crystals.

Table 1 | The summary of 2D Invar and anti-Invar monolayers

Materials	$\alpha (10^{-6} \text{ K}^{-1})$	$E_{2D} (\text{N/m})$	$D (\text{eV})$	$h (\text{\AA})$	type
ZrO ₂	0.47	151.93	3.67	4.96	ZTE
HfO ₂	1.80	157.27	4.48	4.98	ZTE
SbSe ₂	24.49	31.88	9.03	7.13	EPTE
HfBi ₂	16.99	70.99	13.74	9.63	EPTE
HfSb ₂	19.81	55.38	10.01	9.28	EPTE
As ₂	-36.39	51.13	0.81	5.49	ENTE
Ge ₂	-30.46	43.47	0.85	4.88	ENTE
Sn ₂	-61.27	23.96	0.74	5.34	ENTE

The calculated thermal expansion coefficients α (at 500 K, except 40 K for As₂, 90 K for Ge₂, and 80 K for Sn₂), in-plane stiffness E_{2D} , bending stiffness D , and effective thickness h (defines as the geometry thickness, see details in our previous work⁴⁴) for 2D monolayers with ZTE, EPTE and ENTE. The α for ENTE 2D monolayers is the minimal value of $\alpha(T)$ curves.

Discussion

The QHA method employed in our work considers a *free-standing* membrane system with *infinite* size. However, QHA cannot deal with the finite system, systems with boundary conditions (such as substrate effect and local constraints), or the specific entropy-driven wavy conformation. A new theory framework is lacking by now and needs to be established for further research. According to Grüneisen theory, the total thermal expansion can be regarded as the sum of all the possible phonon modes. Because either finite size or boundary conditions would modify the population of phonon modes of the system.

Therefore, the results would be modified²⁸. Because of the weak ability to resist out-of-plane deformation, the surface of 2D crystal would fluctuate at finite temperatures. The values of E_{2D} , D would be influenced by the entropy-driven wavy conformation, which will lead to configuration and temperature-dependent results^{47,48}. In this work, we only consider the values of E_{2D} , D , and α for free-standing 2D crystals with infinite size. And the values of E_{2D} and D are calculated at zero temperature with the flat configuration, which can measure intrinsic properties of bond stretch and out-of-plane deflection. Moreover, these data are easy to be accurately calculate by DFT, while obtaining the temperature-dependent results of E_{2D} and D for extensive materials is challenging. Our results show that descriptors using the data of E_{2D} and D at 0K can well classify PTE and NTE 2D crystals (Fig. 2), which means these data can well capture the thermal expansion at finite temperature. However, it should be noted that its success with free-standing 2D crystals of infinite size does not necessarily guarantee robustness for other systems, such as 2D monolayers with finite size or specific local geometries.

In this work, we identify the interpretable descriptors for the thermal expansion of 2D crystals from extensive data, that is E_{2D} which measures in-plane tensile deformation, and D which measures out-of-plane bending deformation. These two mechanical descriptors can well classify the PTE and NTE 2D crystals. By utilizing high throughput calculations and the state-of-the-art symbolic regression method, these descriptors aid in the discovery of ZTE and ELTE 2D crystals. We discovered the ZTE 2D crystals or 2D Invar materials with very low α in the middle range of temperatures, following the criteria $|\alpha| < 2 \times 10^{-6} \text{ K}^{-1}$. We also discovered EPTE and ENTE 2D crystals or 2D anti-Invar with α larger or smaller than $\pm 15 \times 10^{-6} \text{ K}^{-1}$. The understanding of thermal expansion of 2D crystals, which is dominated by the competition of in-plane and out-of-plane deformation modes, can help us to design ZTE using heterostructure or multi-layer of 2D crystals. For example, the combination of PTE and NTE 2D crystals would lead to ZTE 2D crystals (see Supplementary Fig. 10). Because of different scaling for E_{2D} and D with layer number n ($E_{2D} \sim n^1$, $D \sim n^1$, $\lambda > 1$ ⁴⁶, see details in Supplementary Fig. 11), the multi-layer (or few-layer) 2D crystals would gradually close to ZTE if it is NTE. A good example is graphite that the absolute value of the LTEC is lower than graphene and less than $2 \times 10^{-6} \text{ K}^{-1}$ from 0 and 1000 K¹⁹. In this work, we only consider the macroscopic change of size while that local distortion of lattice, such as rotation, tilting, and collapsing motion, would lead to unusual thermal and mechanical properties^{2,40,49,50}. Consequently, the fine design of local structures, akin to the approaches used in the field of metamaterials, emerges as a promising avenue for exploring unconventional thermal expansion behaviors. Although this work focuses on ZTE 2D crystals, the ELTE 2D materials with large expansion are also valuable for functional designs. For example, the mismatch strain between the PTE and NTE layers can induce the out-of-plane curve deformation if there is no delamination, which is promising for the design of 3D nanostructures with complex geometries^{51,52}. In this work, we exclusively focus on mining 2D monolayers from C2DB. However, further exploration in 2DMatPedia and JARVIS databases^{21,22} could discover new ZTE and ELTE 2D crystals. We expect the discovery of more nanoscale materials exhibiting extraordinary thermal expansion properties that would have profound scientific significance for the design of nano-electronics.

Methods

All DFT computations were performed using the Vienna ab initio Simulation Package^{53,54} based on the projector augmented wave (PAW) method⁵⁵. The generalized gradient approximation of Perdew-Burke-Ernzerhof and the local density approximation exchange-correlation functionals are used^{56,57}. The lattice parameters and atomic coordinates are optimized with the energy convergence of 10^{-8} eV , and the criterion for force convergence during the relaxation is $10^{-6} \text{ eV}/\text{\AA}$. The

cut-off energy for the plane wave is 600 eV, and a mesh grid of k -points $17 \times 17 \times 1$ is adapted to sample the Brillouin zone(BZ). A vacuum space of $\sim 20 \text{ \AA}$ is taken to prevent the interaction between two neighbor layers. The thermal expansion properties of 2D materials are calculated using the quasi-harmonic approximation (QHA) method⁵⁸. The phonon dispersions are obtained by finite difference method using PHONOPY⁵⁹. In this work, we have used a supercell size of $6 \times 6 \times 1$ and a k -point sampling of $4 \times 4 \times 1$. The out-of-plane transverse acoustic (ZA) mode is very soft for 2D materials and a small compressive strain may result in imaginary phonon frequencies near the Γ point. Therefore, small strains (within $\pm 1.5\%$) were employed when utilizing the QHA method. The equilibrium in-plane lattice parameters at any temperature T are determined by minimizing the Helmholtz free energy. Finally, the liner thermal expansion coefficient is defined as $\alpha(T) = \frac{1}{a(T)} \frac{da(T)}{dT}$, where $a(T)$ is the equilibrium lattice constant. In addition to QHA, the LTEC can also be obtained from the Grüneisen theory¹, which shows efficiency and validity (details in Supplementary Fig. 3)^{41,49}. The impact of van der Waals (vdW) dispersion corrections on the LTECs is examined by incorporating DFT-D3 (method of Grimme et al. with zero damping) and vdW-DF vdW dispersion corrections^{60,61}. The results depicted in Supplementary Fig. 12 suggest a negligible influence on the LTECs. For ensure the reliability of our calculation results, we compare our results with available theoretical and experimental data (see details in Supplementary Fig. 13). The results show our calculation results are consistent with reported results and therefore our calculations are reliable. The bending stiffness D can be obtained by fitting the quadratic dispersion $w(q)$ of the flexural ZA mode⁶², $w^2 = \frac{D}{\rho} q^4$, where ρ is the areal density of mass. According to the generalized Hook's law, the in-plane elastic constants can be calculated by applying small strains of $\pm 1\%$ to the equilibrium configuration, and the elastic constants C_{ij} are derived from the second-order derivative of the total energies versus applied strain. And the data of C_{ij} is available in the C2DB. The in-plane stiffness can be derived from the elastic constants as $E_{2D} = (C_{11}C_{22} - C_{12}C_{21})/C_{22}$. The compressibility K_T is calculated as $K_T = S_{11} + S_{12}$. Here, S is the elastic compliance matrix, which can be obtained by calculating the inverse of stiffness matrix C_{ij} .

Data availability

The Source Data underlying the figures of this study are available with the paper. Original data and codes created for the study will also be available on the Materials Cloud platform (<https://doi.org/10.24435/materialscloud:hc-zb>). All the data or scripts generated during the current study are available from the corresponding authors upon request. Source data are provided with this paper.

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Author contributions

K.Z. conceived the project. K.Z. and Y.L. supervised the research. S.T. and K.Z. carried out the DFT calculations and theoretical analysis. S.T., K.Z., W.Y., and Y.L. co-wrote the paper, discussed the results, and commented on the manuscript.

Competing interests

The authors declare no competing interests.

Additional information

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