



NOVEL - MATERIALS  
DISCOVERY LABORATORY



# Find descriptors of Zero-Point-Renormalization(ZPR) by SISSO

Jingkai Quan, coffee talk

25/07/2021  
NOMAD Laboratory

# Content

## ***Introduction:***

- Introduction of Zero-point renormalization
- Introduction of SISSO

## ***Find descriptor of ZPR by SISSO:***

- Focus on stable binary materials
- Including unstable structures
- Next Step

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## ***Introduction:***

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## *Find descriptor of ZPR by SISO:*

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# Zero-point renormalization (ZPR)

□ Electronic structures can be affected by electron-phonon interaction (epi) :

*e.g. temperature dependence of electronic band gap*



Affect band structure even at 0K.

□ Harmonic approximation :

Potential energy Taylor expansion to the displacement  $\Delta\mathbf{R}$  :

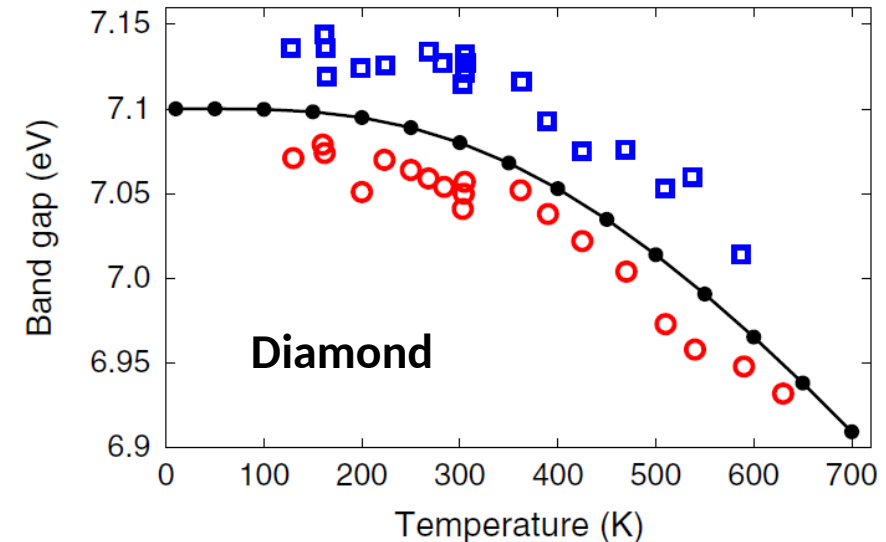
$$E(\{\mathbf{R}^0 + \Delta\mathbf{R}\}) \approx E(\{\mathbf{R}^0\}) + \sum_I \frac{\partial E}{\partial \mathbf{R}_I} \Big|_{\mathbf{R}^0} \Delta\mathbf{R}_I + \frac{1}{2} \sum_{I,J} \frac{\partial^2 E}{\partial \mathbf{R}_I \partial \mathbf{R}_J} \Big|_{\mathbf{R}^0} \Delta\mathbf{R}_I \Delta\mathbf{R}_J + \mathcal{O}(\Delta\mathbf{R}^3)$$

$$\Phi_{IJ} = \frac{\partial^2 E}{\partial \mathbf{R}_I \partial \mathbf{R}_J} \Big|_{\mathbf{R}^0}$$

Hessian matrix (FORCE\_CONSTANT)

DFPT or **Finite difference**

*\*PRL 105, 265501 (2010)*



# Harmonic Approximation

$$\Phi_{IJ} = \left. \frac{\partial^2 E}{\partial \mathbf{R}_I \partial \mathbf{R}_J} \right|_{\mathbf{R}^0} \longrightarrow \mathbf{F}_I = - \sum_J \Phi_{IJ} \Delta \mathbf{R}_J$$

$$\begin{pmatrix} m_1 \Delta \ddot{\mathbf{R}}_1 \\ m_2 \Delta \ddot{\mathbf{R}}_2 \\ \vdots \\ m_N \Delta \ddot{\mathbf{R}}_N \end{pmatrix} = \begin{pmatrix} \Phi_{11} & \Phi_{12} & \cdots & \Phi_{1N} \\ \Phi_{21} & \Phi_{22} & \cdots & \Phi_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ \Phi_{N1} & \Phi_{N2} & \cdots & \Phi_{NN} \end{pmatrix} \begin{pmatrix} \Delta \mathbf{R}_1 \\ \Delta \mathbf{R}_2 \\ \vdots \\ \Delta \mathbf{R}_N \end{pmatrix}$$

↓ **Fourier**

**Eigenvalue Problem:**

$$D(\mathbf{q}) \mathbf{e}_s(\mathbf{q}) = \omega_s^2(\mathbf{q}) \mathbf{e}_s(\mathbf{q})$$

$\mathbf{e}_s(\mathbf{q})$  : **eigenvector (displacement)**

$$D_{IJ}(\mathbf{q}) = \sum_{J'} \frac{e^{i(\mathbf{q} \cdot \mathbf{R}_{JJ'})}}{\sqrt{M_I M_J}} \Phi_{IJ'}$$

**Dynamic matrix (fourier trans. of  $\Phi$ )**

tip1: differential property of fourier trans.

$$f^{(n)}(x) \text{ Fourier } i\omega^n F(\omega)$$

$$f^{(2)}(x) \text{ Fourier } -\omega^2 F(\omega)$$

↔

# Zero-point renormalization (ZPR)

Total zero-point renormalization:

$$\delta\varepsilon_i(\mathbf{k}) = \delta\varepsilon_i^A(\mathbf{k}) + \delta\varepsilon_i^P(\mathbf{k})$$

$\delta\varepsilon_i^A(\mathbf{k})$  : Adiabatic term

$\delta\varepsilon_i^P(\mathbf{k})$  : polar term

*\*J. Phys. Chem. C 2021, 125, 6479–6485*

*\*Phys. Rev. Lett. 112, 215501*

*\*Phys. Rev. Lett. 94, 036801*

□ Adiabatic term:

$$D(\mathbf{q}) e_s(\mathbf{q}) = \omega_s^2(\mathbf{q}) e_s(\mathbf{q})$$

$e_s(\mathbf{q})$  : eigenvector (displacement)

$$\delta\varepsilon_i^A(\mathbf{k}) = \sum_{\lambda} \int \frac{d\mathbf{q}}{\Omega_{\text{BZ}}} \frac{\hbar}{4\omega_{\lambda}(\mathbf{q})} \frac{\partial^2 \varepsilon_i(\mathbf{k})}{\partial \delta_{\lambda q}^2}$$

Expand by eigen displacements

$$\delta\varepsilon_i^A(k) = \sum_{\lambda} \int \frac{d\mathbf{q}}{\Omega_{\text{BZ}}} \frac{\hbar}{4\omega_{\lambda}(\mathbf{q})} \frac{e_{I\alpha,\lambda}(\mathbf{q})}{\sqrt{M_I}} \frac{\partial^2 \varepsilon_i(\mathbf{k})}{\partial R_{I\alpha} \partial R_{J\beta}} \frac{e_{J\beta,\lambda}(\mathbf{q})}{\sqrt{M_J}}$$

$$\delta\varepsilon_i^A(\mathbf{k}) \quad \frac{1}{(M_I \cdot M_J)^{1/4}}$$

# Zero-point renormalization (ZPR)

□ Polar term: Fröhlich model by Nery and Allen

\**J. Phys. Chem. C* 2021, 125, 6479–6485

\**Phys. Rev. B* 94, 115135

$$\delta\varepsilon_i^{\text{P}}(\mathbf{k}) = \pm \frac{\alpha_i \hbar \omega_{\text{LO}}}{\pi} \tan^{-1} \left( \frac{q_{\text{F}}}{q_i} \right)$$

$$\alpha_i = \frac{e^2 q_i}{8\pi \tilde{\varepsilon}_0 \hbar \omega_{\text{LO}}} \left( \frac{1}{\varepsilon_{\infty}} - \frac{1}{\varepsilon_0} \right)$$

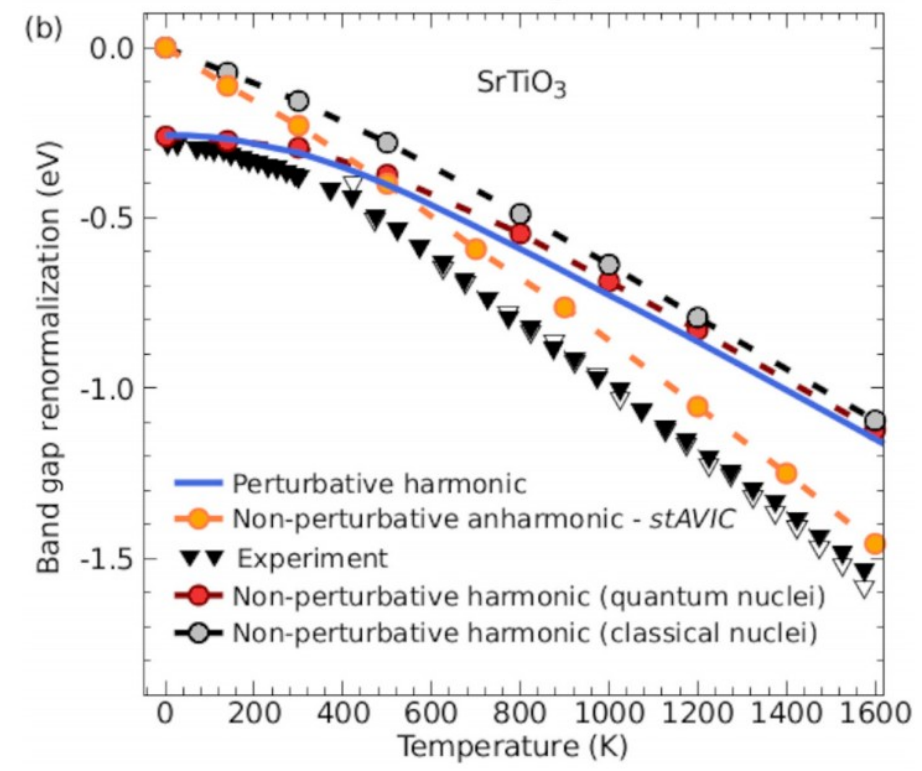
$$q_i = \sqrt{\frac{2m_i^* \omega_{\text{LO}}}{\hbar}}$$

$$\delta\varepsilon_i^{\text{A}}(\mathbf{k}) \quad \omega_{\text{LO}}$$

$$\delta\varepsilon_i^{\text{A}}(\mathbf{k}) \quad m^*$$

# Beyond harmonic approximate

- There also beyond the harmonic/perturbative pictures exist  
but that they are not used here.



\*Zacharias, M., Scheffler, M. & Carbogno, C. *Phys. Rev. B* 102, 045126

# Content

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- Introduction of SISO

## ***Find descriptor of ZPR by SISO:***

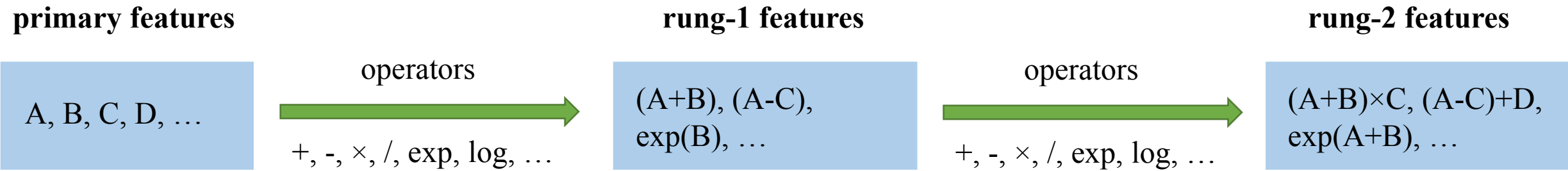
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# SISSO: a compressed-sensing method

## □ Sure Independence Screening (SIS) + Sparsifying Operator (SO)

\* *Phys. Rev. Materials* 2, 083802

STEP 1. Construct large feature space (hyperparameter: **rung = n**)

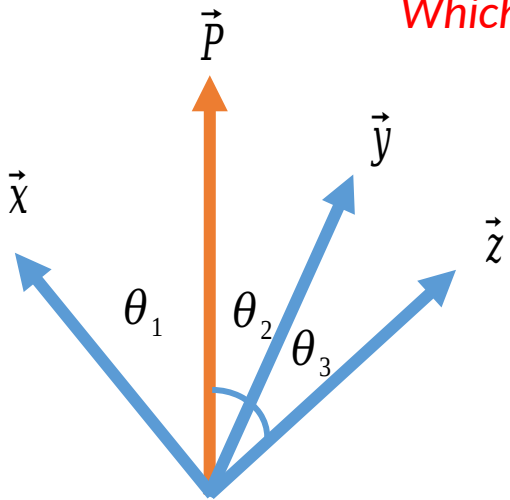


\* Size of feature space grow rapidly with each iteration

# SISSO: a compressed-sensing method

STEP 2. SIS: reduce feature space size by pearson correlation

Which vector most correlated with target vector P?



- **Pearson correlation:** First **zero-centering** vectors, then calculate cosine

$$corr(x, P) = \frac{\sum_s (x_s - \bar{x})(P_s - \bar{P})}{\sqrt{\sum_s (x_s - \bar{x})^2} \sqrt{\sum_s (P_s - \bar{P})^2}}$$

$$\cos \theta = \frac{\sum_s (x_s P_s)}{\sqrt{\sum_s x_s^2} \sqrt{\sum_s P_s^2}}$$

$$\vec{x} = (x_1, x_2, \dots, x_s)$$

$$\bar{x} = \frac{\sum_s x_s}{s}$$

We can select top ranked features to form a feature subspace (hyperparameter:  $n_{sis\_select} = n$ )

e.g. choose 10000 top ranked features out of feature space.

# SISSO: a compressed-sensing method

STEP 3. SO: Find desired descriptor in smaller feature subspace (  $l_0$ -norm)

Find the sparse solution  $c$  in feature subspace: **minimization problem**

$$\underset{c}{\operatorname{argmin}} (\|y - Dc\|_2^2 + \lambda \|c\|_0)$$

$\|c\|_0$  = number of nonzero components of  $c$

MT-SISSO (Multi Task SISSO):

Can find an overall descriptor for multiple different but related properties. (only coefficients are different)

\* *Phys. Rev. Materials* **2**, 083802

*J. Phys. Mater.* **2** 024002

# Content

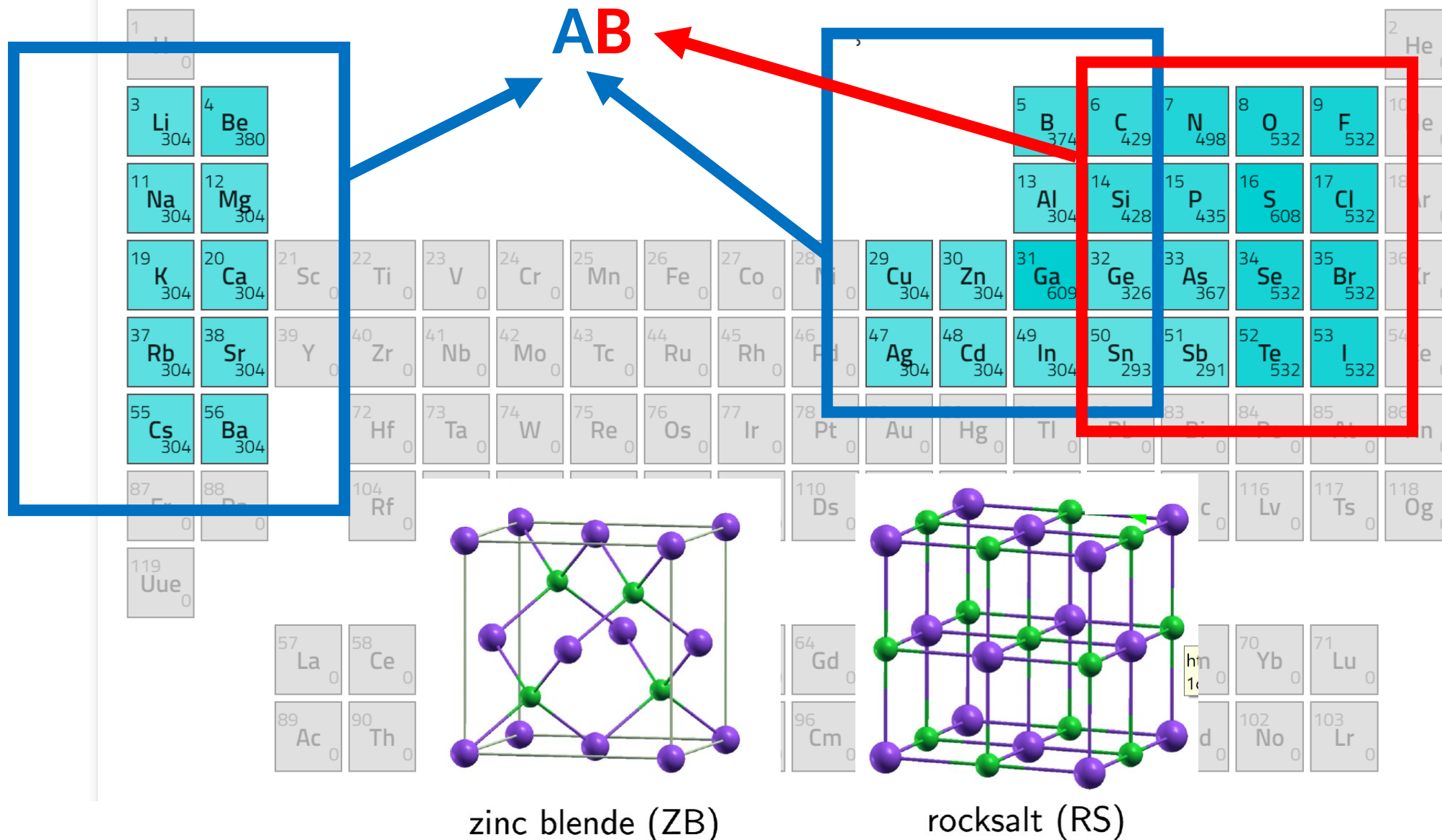
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# Overview: ZPR Dataset of binary materials



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# Focus on stable structures

1. **82 rocksalt (RS) and 82 zincblende (ZB) structures in total.**

2. **stable structures (total energy lowest) and leave metals out**

- RS structure: 38 materials
- ZB structure: 37 materials

3. **direct gap ZPR at  $\Gamma$ -point data is available with and w/o polar correlation term.**

**ZPR values calculated by Honghui, also uploaded on NOMAD.**

*\*J. Phys. Chem. C 2021, 125, 6479–6485*

# Primary Features

□ Property to learn: **ZPR at  $\Gamma$  point**

**30** primary features:

- **Effective mass at  $\Gamma$  point:**
- **Nuclear Mass related:**
- **Atomic features:**

$$*mass\ factor = \frac{1}{(mass_A \cdot mass_B)}$$

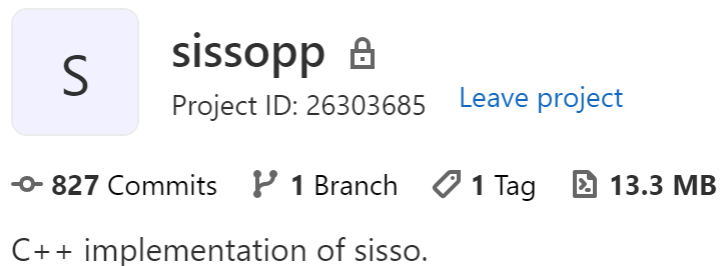
*(IP=ionization potential, EA=electron affinity,  $r_s$ =radius of s orbital,  $r_{val}$ =radius of valance orbital)*



- **Binary features:**,  
*(dis=nearest atom distance, bond=bond length of AB, be=bond energy, hlg=Homo-Lumo gap)*
- **El-Ph-Related features:**





# Softwares and codes

## 1. C++ implementation of **SISSO** developed by Tom

sissopp\_developers > sissopp




**sissopp**   
 Project ID: 26303685 [Leave project](#)

 827 Commits  1 Branch  1 Tag  13.3 MB

C++ implementation of sisso.

[\\*https://gitlab.com/sissopp\\_developers/sissopp](https://gitlab.com/sissopp_developers/sissopp)

## 2. **FHI-aims** for structure relaxation and band structure calculation



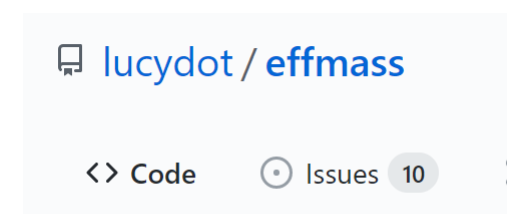
*\*Computer Physics Communications* **180**, 2175-2196 (2009)

## 3. **FHI-vibes** for phonon dispersion calculation



*\*Journal of Open Source Software*, 5(56), 2671

## 4. **effmass** for effective mass calculation



[\\*https://github.com/lucydot/effmass](https://github.com/lucydot/effmass)

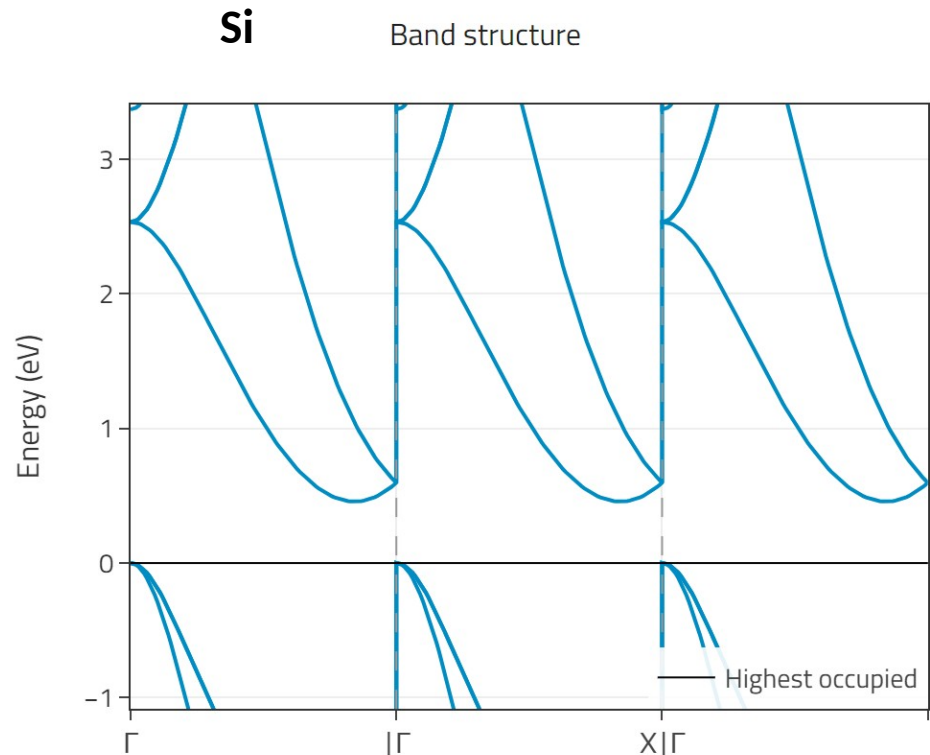
# Effective masses at $\Gamma$ point

$$\frac{1}{m^*} = \frac{1}{\hbar^2} \frac{\partial^2 E}{\partial k^2}$$

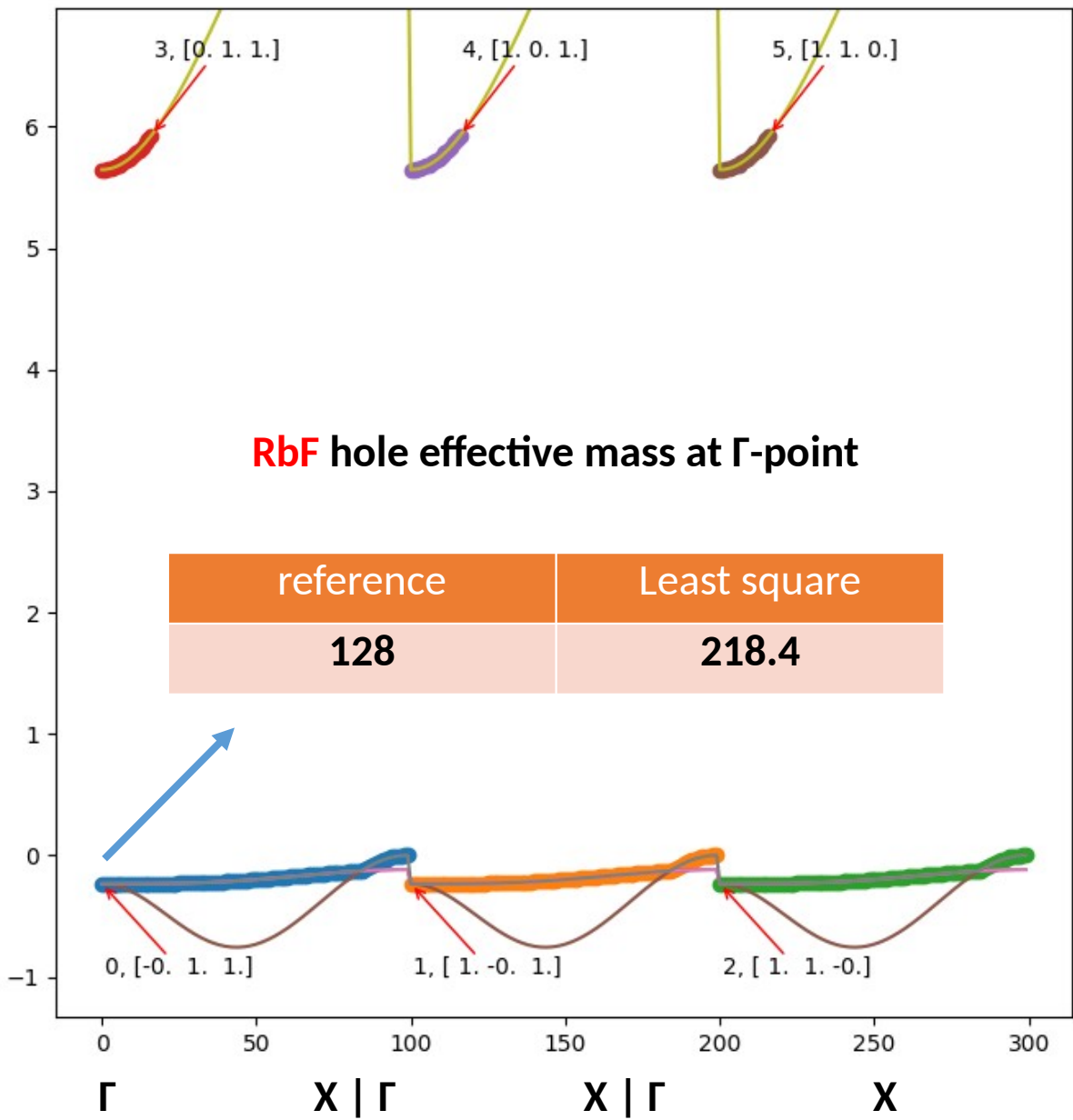
Least square method : parabolic approximate

It's usually a good approximation, but:

**Don't work for linear band dispersion!**



# Effective masses at $\Gamma$ point



\*Finite difference method :

reference	Finite difference
128	102.776

- Thus use finite difference method to calculate effective masses by *effmass* code.

\*<https://github.com/lucydot/effmass>

\*Phys. Rev. B 99(8), 085207

# SISSO on stable structures

1. Use SISSO and **Multitask**-SISSO to find descriptors
2. Performed 50 **cross-validation (leave-10%-out)** on less primary features

#primary features	30	14
Time spent	~20h (Draco); ~8h (Raven)	~15min

**Hyper-parameters:**

- `rung = 3`
- `n_sis_select = 10,000`

(Tested `n_sis_select = 5000`; `10,000` and `50,000`. They all find same descriptors.)

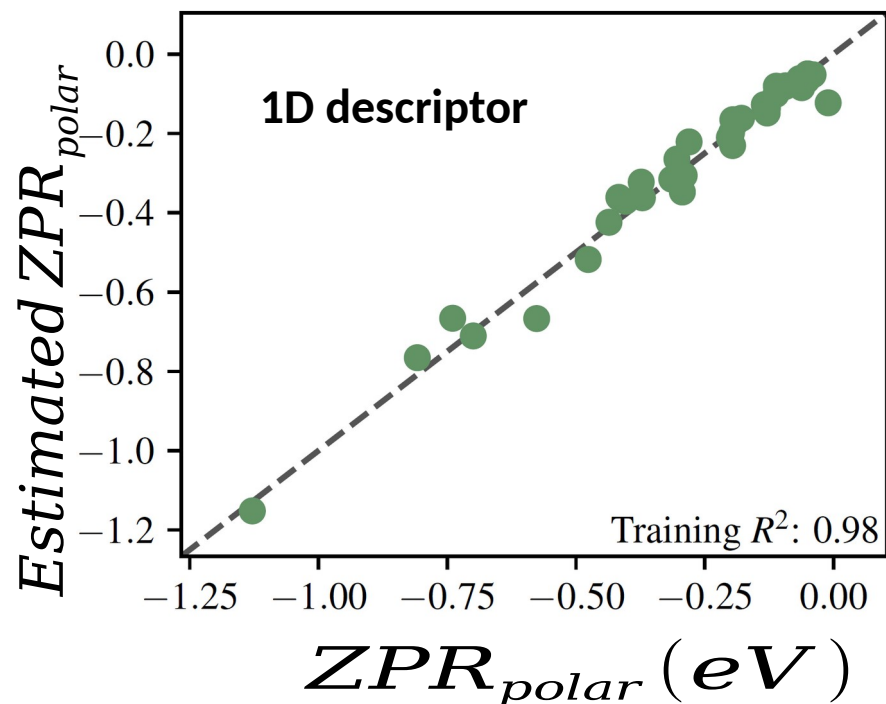
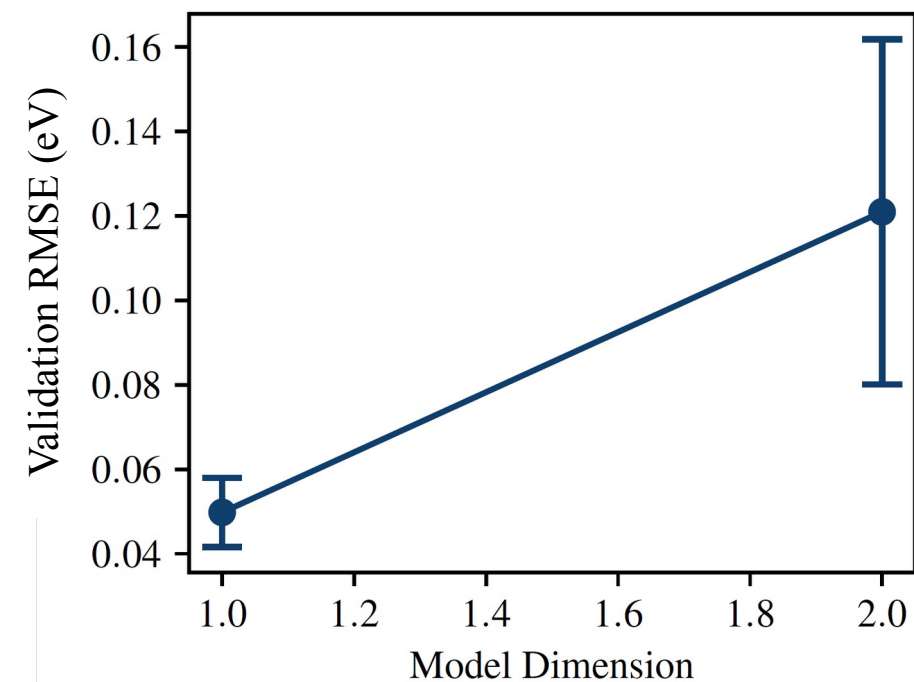


# Cross-Validation for RS structure

mean RMSE on 50 leave-10%-out cross validations

- rung = 3
- n\_sis\_select = 10,000

*Same parameters as our regression.*



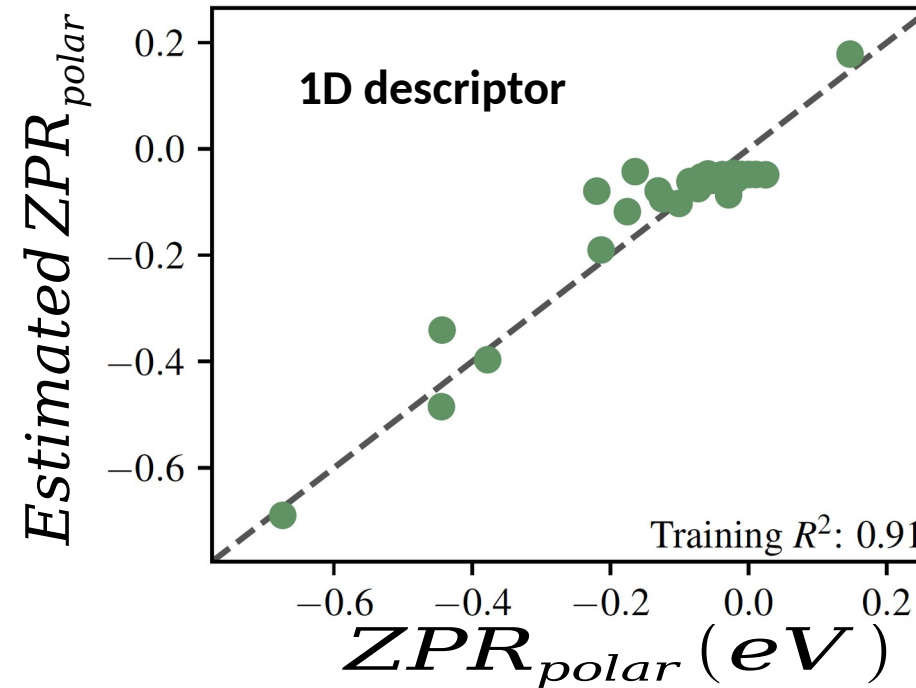
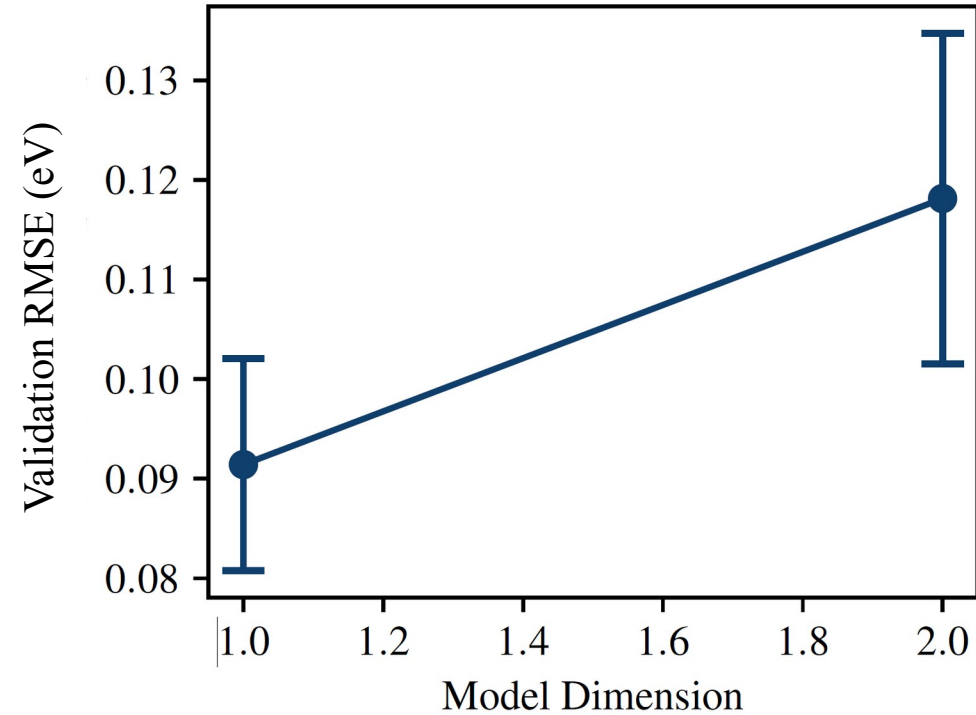
$$1D \text{ descriptor} = c_0 + a_0 \cdot \frac{\sqrt[3]{mass_{electron} \cdot beBB \cdot homo_B \cdot IP_B}}{dis_{AB} \cdot \sqrt[3]{IP_A}}$$

# Cross-Validation for ZB structure

mean RMSE on 50 leave-10%-out cross validations

- rung = 3
- n\_sis\_select = 10,000

*Same parameters as our regression.*



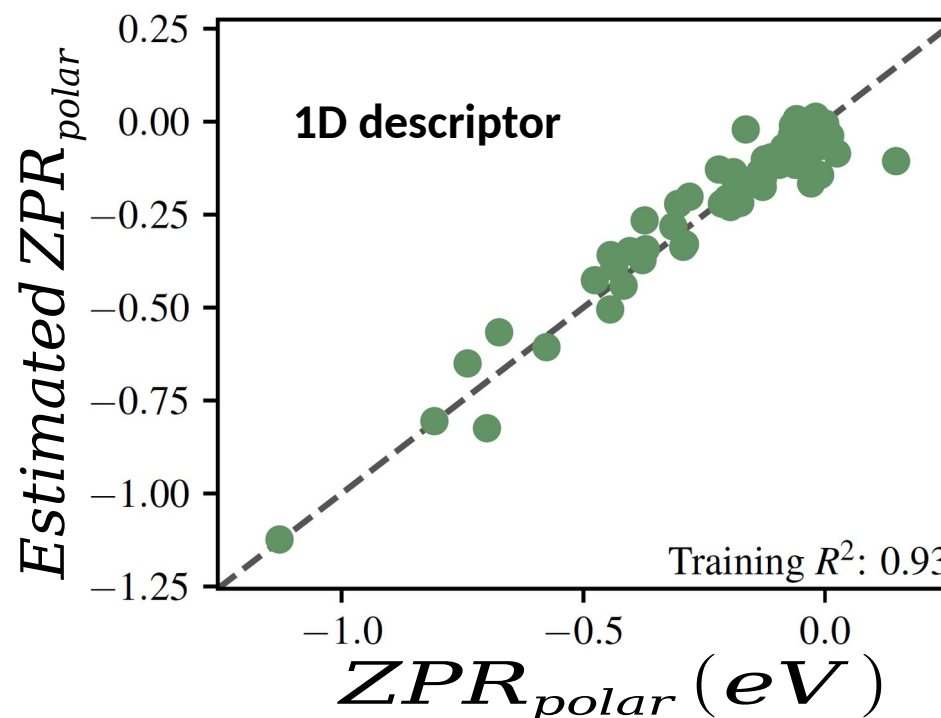
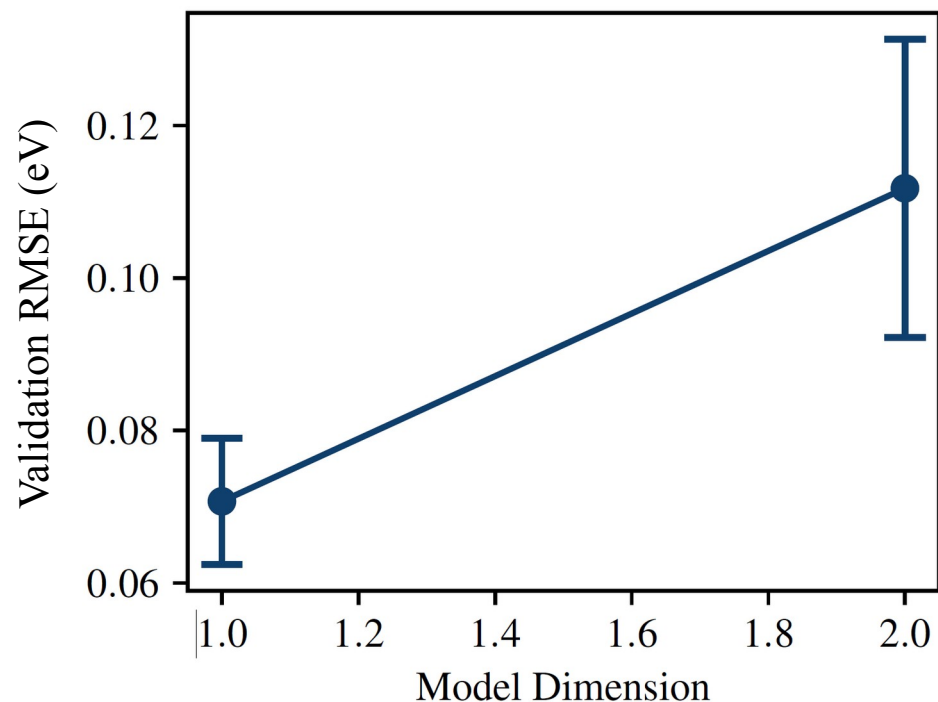
1D descriptor:  $D = c_0 + a_0 \cdot \frac{mass_{hole} \cdot be_{BB} \cdot be_{AB}}{mass_B \cdot (IP_A \cdot mass_B + IP_B \cdot mass_A)}$

# Cross-Validation for Multi-Task SISSO

mean RMSE on 50 leave-10%-out cross validations

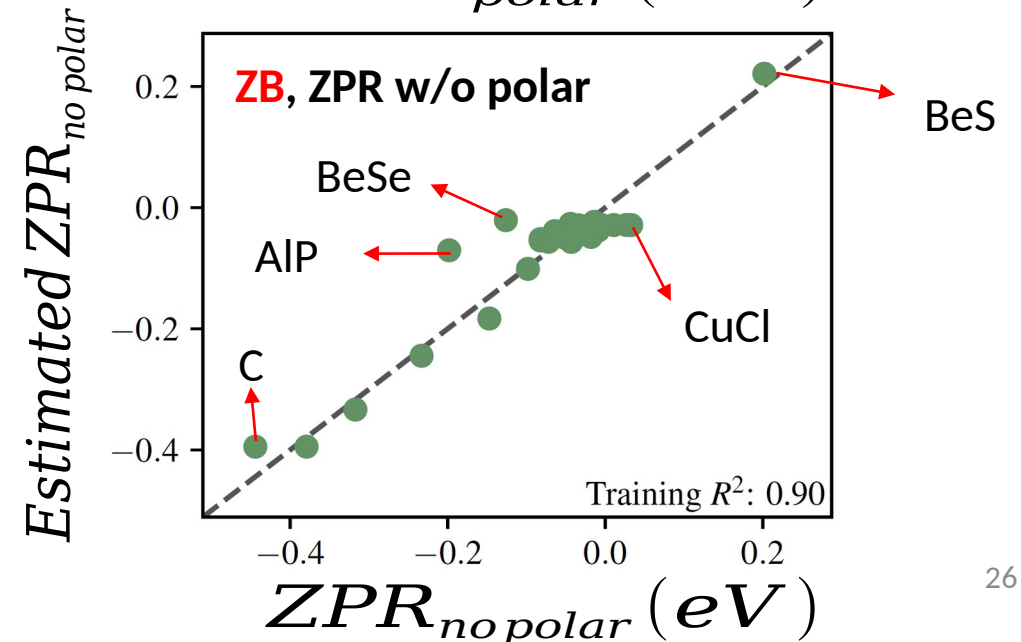
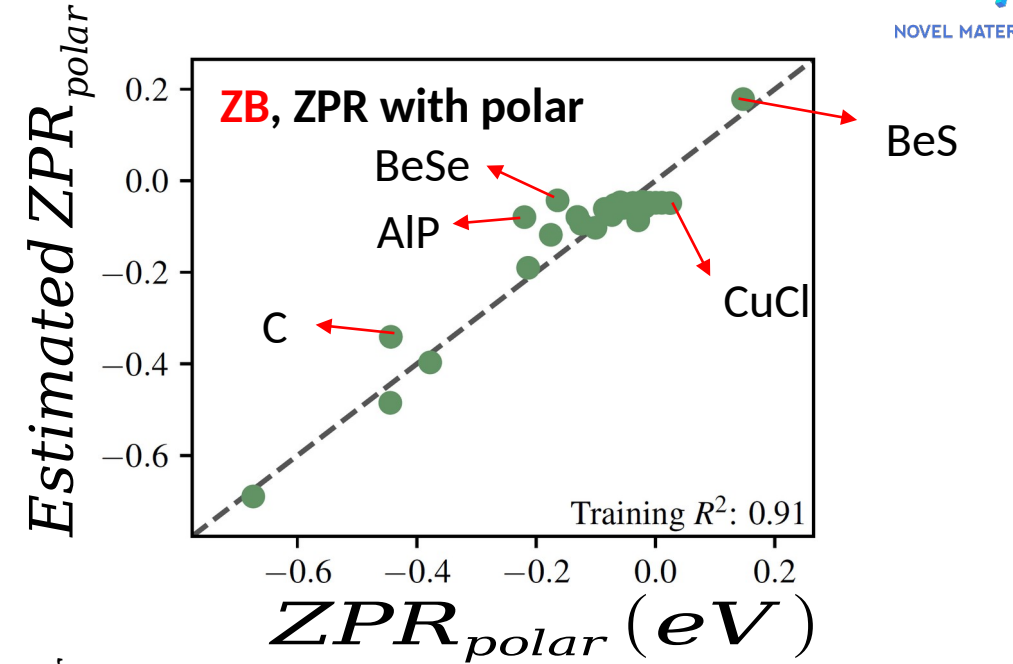
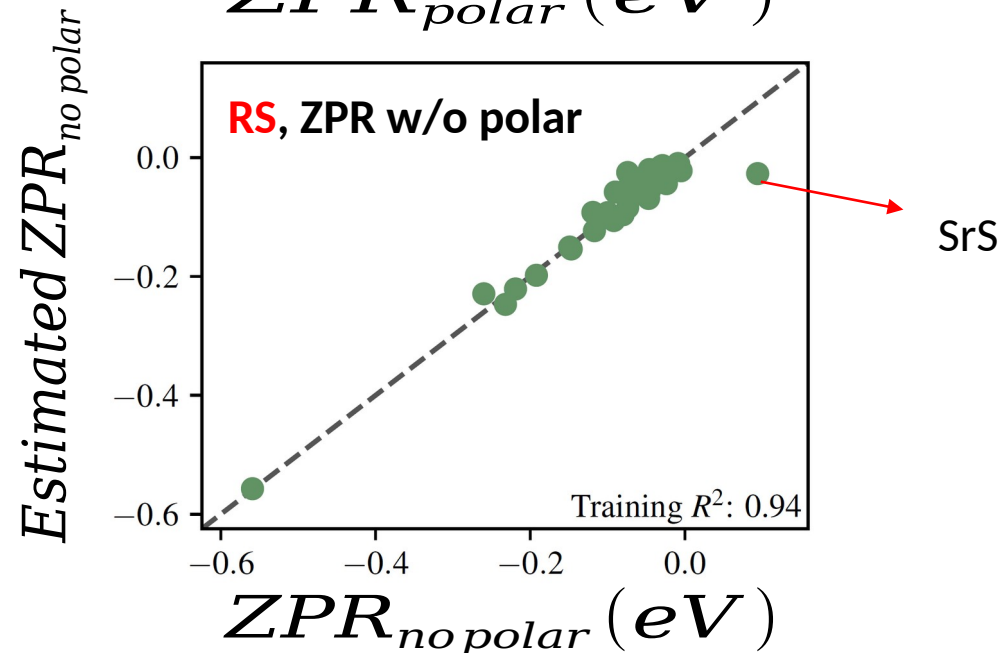
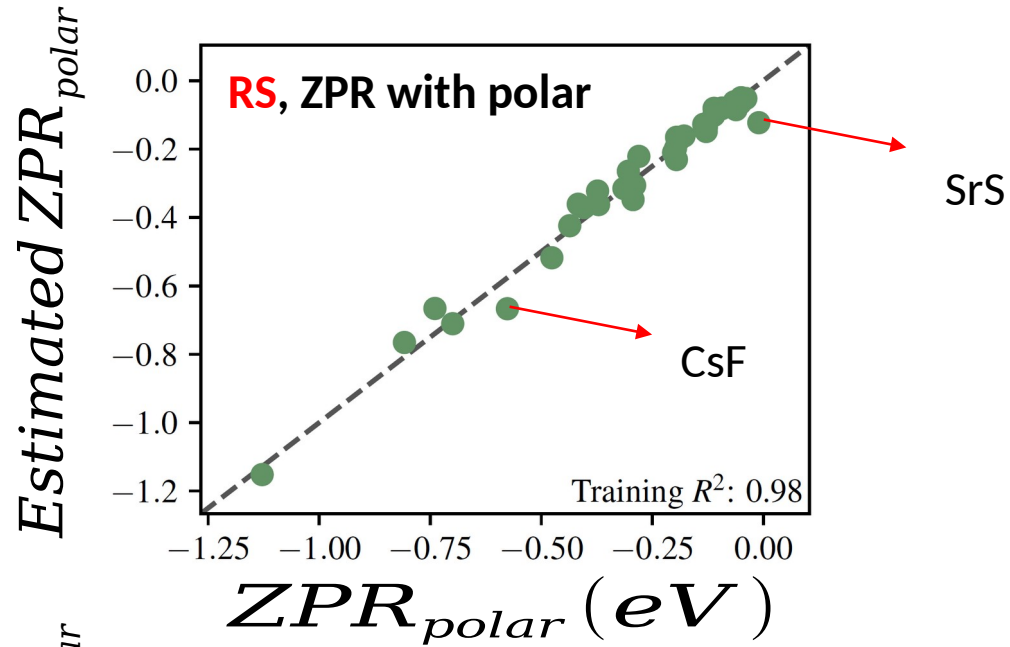
- rung = 3
- n\_sis\_select = 10,000

*Same parameters as our regression.*



$$1D \text{ descriptor: } \Delta_{a-a} = c_0 + a_0 \cdot \frac{\Delta_{a-a}^3 + \frac{homo_B}{homo_A}}{\sqrt[3]{average_{mass} \cdot r_{val_B} \cdot homo_B}}$$

# Including metastable structures



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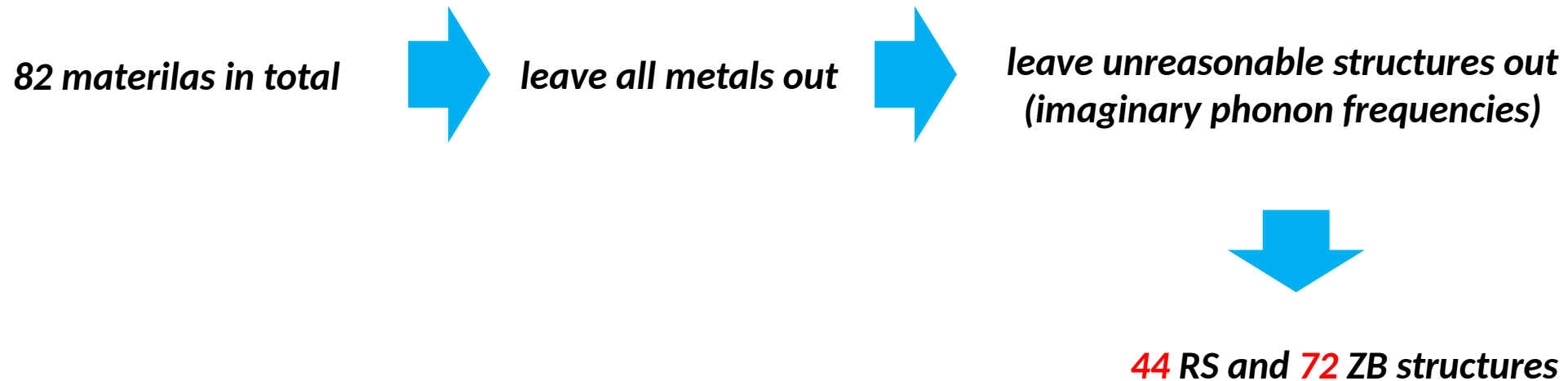
## *Find descriptor of ZPR by SISO:*

- Focus on stable binary materials
- Including metastable structures
- Next Step

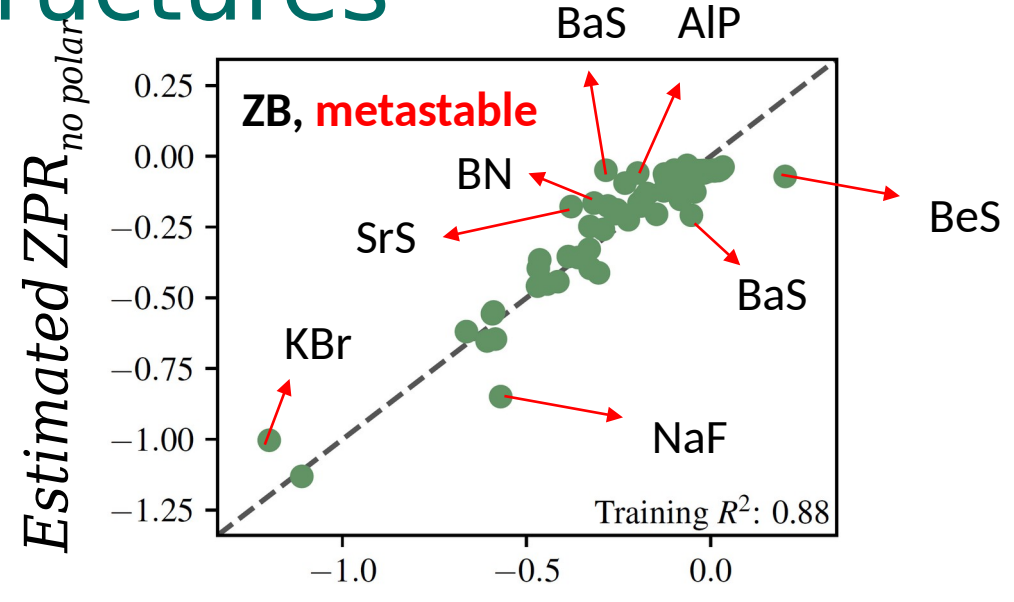
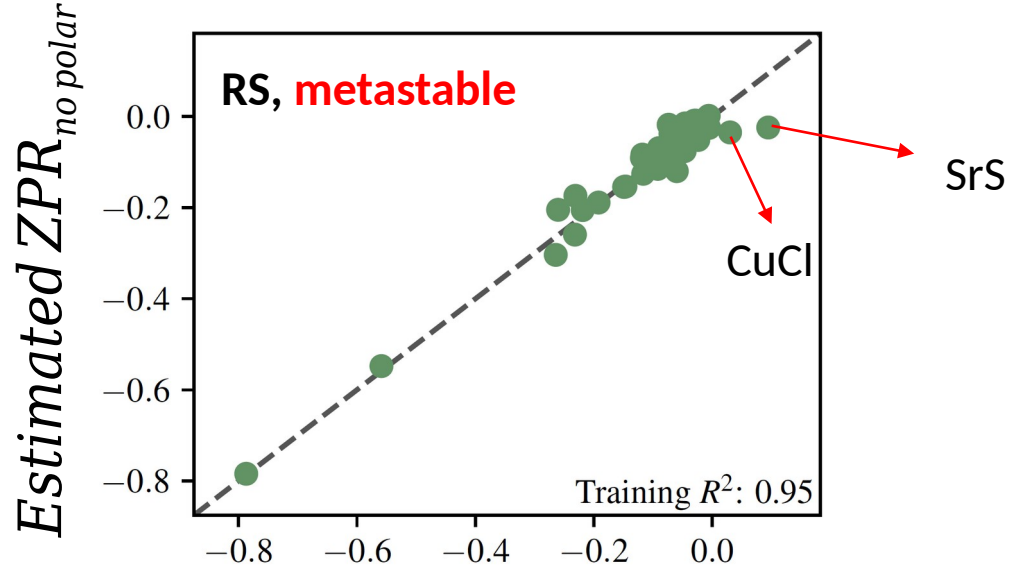
# Including metastable structures

□ More data points can help find better descriptor

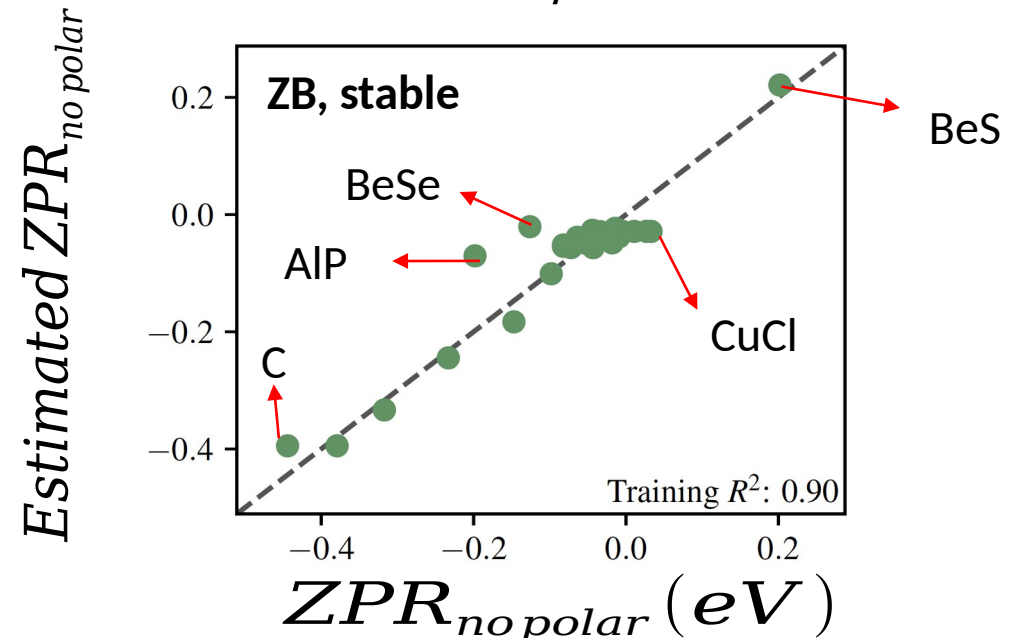
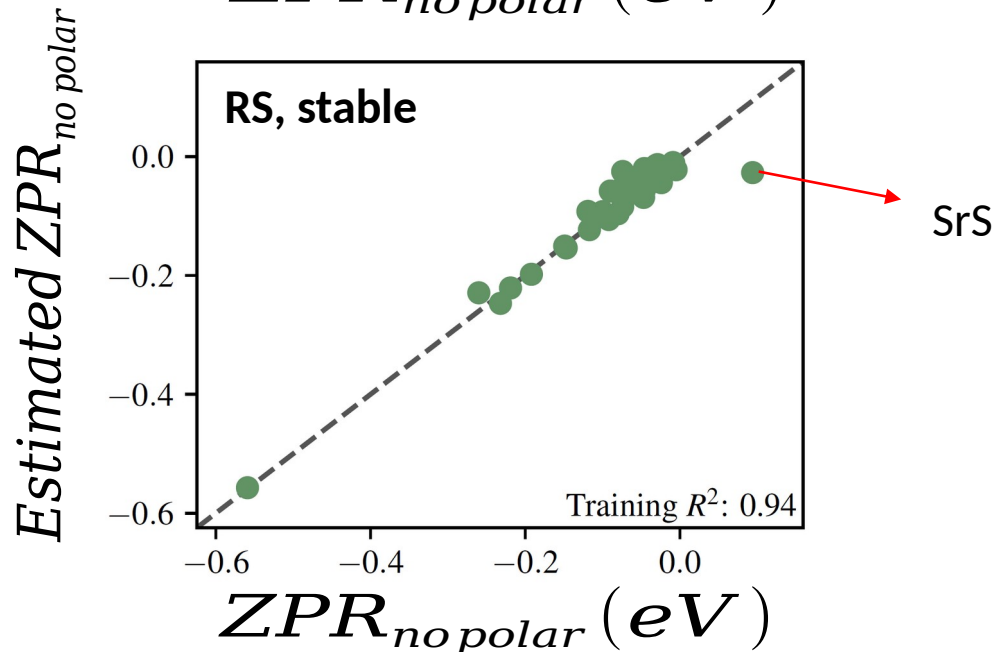
\* **Metastable** means: 1. total energy a little higher than most stable structures.  
2. Have no imaginary phonon frequencies.



# SISSO on metastable structures



*Haven't done cross-validation yet*



# Next Steps

- *Further Cross-validation of descriptors for metastable structures*
- *Investigate [correctness of] outliers and physical effects leading to a different behaviour.*
- *Compare with experimental and theoretical literature results [only few data available, unfortunately!] and add them to the learning set.*

Thank you !